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Contributions to the assessment of demersal fish stocks off Namibia

Johannes Nduvudi Kathena

PhD Thesis
Contributions to the assessment of demersal fish stocks off Namibia

PhD thesis by:
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June 2019

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Preface

This thesis was prepared at the Technical University of Denmark (DTU) National Institute of Aquatic Resources (DTU Aqua), under supervision of Associate professor Uffe Høgsbro Thygesen (DTU Aqua, DTU Compute since June 2017), Senior Researchers Anders Nielsen and Teunis Jansen and Hashali Hamukwaya (Benguela Current Convention, BCC). The project was funded by the Development of ecological sustainable fisheries practices in the Benguela Current Large Marine Ecosystem (ECOFISH).

This thesis consists of a series of 4 scientific papers.

Kongens Lyngby, June 2019

Johannes Nduvudi Kathena
Dedication

This thesis is dedicated to my late mother: Naango, late father: Ignatius and late father-in-law: Jason. Their love for education has led me to pursue a Ph.D. I am deeply grateful for their support and influence during my battle of pursuing school through to a Ph.D. It saddens me greatly that they did not see my battle through to the end of my formal qualifications. May their souls continue to rest in eternal peace!
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Special thanks go to my mentor Jan E. Beyer. Jan has continuously encouraged me throughout this project. He has been very supportive and full of great ideas. Jan was always there to share and discuss ideas about stock assessment ranging from classical assumptions of fishery biology to state-space modelling. You have truly been an amazing mentor.

Thank you to the Benguela Current Convention ECOFISH project (funded by the European Union under the Europe Aid grant 2010/222387) for providing the funds necessary for undertaking this Ph.D. I am also grateful to the Ministry of Fisheries and Marine Resource, Namibia, and National Institute of Aquatic Resources, DTU for the study opportunity. Thank you moreover to Jan E. Beyer, Hashali Hamukuaya and Fritz Köster without whom I might not have been offered the opportunity to do this Ph.D. in the first place.

I wish to acknowledge and honour the contributions of my research associates: Jan E. Beyer, Alexandros Kokkalis, Casper Willestofte Berg, Martin W. Pedersen, Espen Johnsen, Dawit Yemane and Nixon Bahamon for their various contributions to this project.

Many people have contributed to the realization of this thesis in different ways and I do thank them wholeheartedly: Lillian Andersen, Nina Qvistgaard, Eva Maria Pedersen, Lis Ramberg, Alexandros Kokkalis,

I had the privilege to share an office with the modelling group, SAM and SPiCT developer, thank you all for the support and encouragement you have offered along the journey that has been the undertaking of this Ph.D. Without you, I would not have discovered the concepts (“random walk and variance parameters in fisheries modelling”). I am very appreciative of your significant contribution to the work involved in developing the two models that are described and explored in this thesis.

Motsazizowala - your reliable support and encouragement are highly appreciated. I look forward to seeing more of such projects.

Most important of course my family: Tulonga, Tangomuwa, Taamba and my lovely wife Sabina, I appreciate all of you in every sense of the word. Thank you for your support at all levels, it kept me going. I am sorry that I was not always home when you needed me the most, but that is for a good reason.
Summary (English)

In Northern Benguela, the two hake species *M. capensis* and *M. paradoxus* with similar visual appearance have a common habitat and overlapping depth and latitudinal distribution. These species can be found in the same hauls in what can be referred to as a mixed fishery. The challenges in mixed fisheries are that it is often not possible to obtain Maximum Sustainable Yield from all stocks at the same time. Species-specific assessments are important for management and sustainable utilization of the fisheries resources. Since the two species are similar and not distinguished by the commercial fisheries, methods should be developed to separate the species in commercial landings.

This thesis has developed splitting algorithms to split the two hake species in the commercial landings using observer and survey data (Paper I). The results support the previous findings: *M. capensis* had a shallower and more northern distribution than *M. paradoxus,* while *M. paradoxus* dominated the annual hake catches. Based on the results from Paper I, the thesis proceeds to develop a species-specific abundance at each commercial haul (Paper II). We develop a generalized additive model (GAM) to identify factors that are affecting catch rates or catch-per-unit-of-effort (CPUE) of *M. capensis* and *M. paradoxus* for each tow (Paper II). Standardized catch rates for *M. paradoxus* are significantly higher than catch rates of *M. capensis.* Furthermore, seasonal migration patterns are described for the first time and are found to correspond to the spawning areas and season. We conclude on hake species-specific studies by implementing a data-rich state-space assessment model (SAM) (Paper III) for two independent single-species assessments, one for *M. capensis,* and the other for *M. paradoxus.* *M. capensis* showed a higher spawning stock biomass than *M. paradoxus,* while fishing mortality was higher on *M. paradoxus.*

Finally, in (Paper IV) a biomass dynamic model (SPiCT) was applied to a data-rich Cape monkfish and a data-moderate West Coast sole with a perspective of evaluating their use in Northern Benguela with a particular focus on assessing demersal resources off Namibia.

Overall, the thesis work has constructed state-of-the-art single-species assessment models of hake and monkfish, using SAM, has explored the applicability of data-moderate assessment methods, specifically
SPiCT, to monkfish and applied it to west coast sole, and has examined the differences in distribution between the two hake species, thus addressing the species splitting problem which is a prerequisite for separate assessments of the two species. Thus, the work has contributed to the advance of the state-of-the-art with respect to fisheries assessment and sustainable management of the living resources off Namibia and in the Benguela current system.
Summary (Danish)

De to kulmule-arter, *M. capensis* og *M. paradoxus* i det nordlige Benguela, har meget ens udseende, fælles habitat og overlappende dybde og breddegrads-fordeling. Disse arter kan findes i samme trawltræk i, hvad der kan betegnes som et blandet fiskeri. Udfordringerne i blandet fiskeri er, at det ofte ikke er muligt at opnå maksimalt bæredygtige fangster (Maximum Sustainable Yields) fra alle bestande på samme tid. Separate bestandsvurderinger for hver art er vigtige for management og bæredygtig udnyttelse af fiskeri ressourcer. Siden de to arter ligner hinanden, og det kommercielle fiskeri ikke skelner mellem dem, bør der udvikles metoder til at adskille arterne i de kommercielle landinger.

Til slut (Paper IV) blev en dynamisk biomasse model (SPiCT) anvendt på en data-rig Cape havtaskebestand og en data-moderat bestand af Vestkyst-tunge med det perspektiv at evaluere deres nytte i det nordlige Benguela, med særlig fokus på at bestandsvurdere Namibias demersale ressourcer.

Samlet har arbejdet med afhandlingen skabt state-of-the-art enkeltart bestandsvurderings modeller for kulmule og havtaske ved at anvende SAM har udforsket anvendeligheden af data-moderate bestandsvurderingsmetoder, specielt SPiCT, for havtaske samt anvendt den på vestkyst-tunge, og har undersøgt forskellene i fordeling mellem de to kulmulearter, og således arbejdet med ”the species splitting problem”, som er en forudsætning for seperate bestandsvurderinger af de to arter. Arbejdet har således bidraget til at fremme state-of-the-art hvad angår fiskerivurderinger og bæredygtig management af de levende fiskeri-ressourcer i Namibia og i Benguela-strøm-systemet.
List of publications

The thesis is based on the following original research papers:


The following list of publications presents co-authored papers written during the PhD project which are either bordering on the subject in this thesis, or where the author of this thesis did not have a leading role in producing the manuscript, or a combination of the two. The content of these papers will not be discussed directly, but are included in the discussions of the main papers where relevant.


Chapter 1: Introduction

1.1 Motivation

In 2012, at the International Stock Assessment Workshop at the University of Cape Town, South Africa, Danish fisheries scientist Dr Anders Nielsen stunned the participants with a presentation on a state-of-the-art stock assessment model (SAM). A state-space model that can be executed on a web interface so assessments can be conducted simultaneously by many people in different places of the world. The presentation and the subsequent discussion on SAM triggered my appetite to pursue this field. For biologists, the innovation of assessing a stock through a web interface has made it easier to conduct assessments without going into the technical details of the machine room.

This experience motivated me to study SAM and similar stochastic assessment techniques. However, my aim should not be to contribute to statistic stock assessment methodology but rather to demonstrate applications of such methodology. This would help Namibian fishery biologists continue this new line of implementing and interpreting state-of-the-art assessment techniques.

The focus of my work has been Cape hakes, Cape monkfish and West Coach sole, the latter as a representative of the data-poor stocks not yet assessed in Namibia.

Cape hake consists mainly of two species which in Namibia today are assessed and managed as one combined species. Since hake in every respect constitutes the most important fishery in Namibia, I considered it a great motivation to investigate how available knowledge from commercial catch rates together with species-specific survey information could be used to provide new input for stock assessment allowing for separate assessments of the two hake species. It is from this basis with SAM as a starting point that I developed clear objectives for my PhD study with an ambition of producing a thesis that would advance the field of science and critically discuss the results.
1.2 Objectives

The overall objectives of this PhD study were to produce a single-species assessment of four demersal fish stocks off Namibia. Specifically:

- Develop a splitting method for the commercial catches of the two hake species.
- Use the time-series of the separated hake commercial catches to:
  - Extract information from commercial CPUE to infer population abundance and seasonal migration patterns.
  - Use the same data-rich state-space assessment modelling for comparing combined hake assessment with individual hake species assessment.
- Compare results of data-rich to the data-moderate state-space assessment of monkfish.
- Perform a data-moderate state-space assessment of west coast sole.
- Disseminate all results to the scientific community, in particularly in the region, with the ultimate goal of facilitating improved assessment, advice, and management in the Benguela Current Large Marine Ecosystem.

1.3 Overview

This thesis uses data of four demersal living resources off Namibia with which I examine a series of issues related to improving stock assessment of these resources. Although the results are specific to these species, the assessment methods and approaches are of general relevance to many of the assessed and/or unassessed stocks off Namibia, in the Benguela current system and elsewhere. I assess these stocks using the established assessment methods, mainly SAM and SPiCT, currently used in ICES. These established methods are used to perform an assessment for data-rich and/or data-moderate stocks. I describe the dynamics of these fish stocks, present the results and discuss the new perceptions.

Chapter 2 provides background information on the Benguela Current Large Marine Ecosystem and summarizes the biology and ecology of the four species, the fisheries that target them, monitoring programs,
assessment and management. The information is mainly gathered from the literature. It is of relevance to the understanding of the role these living resources play in the ecosystem. They also provide input data to the models used in the thesis. Finally the chapter highlights three current trends in fisheries stock assessment which formed the background and inspiration for the current thesis: First, the increased interest in data limited assessment methods; second, a more rigorous application of statistical methods in assessment, e.g. to quantify confidence; and third, emphasizing transparency in model coding and model formulation.

Chapter 3 presents the contribution of the thesis. The original research is documented in papers which are in appendices, but this chapter summarizes the findings, adds some extra content, discussion and extensions. In Paper I, we developed the algorithm of splitting commercial catches into hake species-specific ratio. The algorithm was presented at the International stock assessment review panel at the University of Cape Town, Cape Town, South Africa in December 2011. Critical review and input by the panel are incorporated in the algorithm. In Paper II, we proceeded by applying GAMs to the data source of haul-by-haul species-specific commercial catch ratio. In Paper III, we implemented a species-specific assessment using the state-space assessment model (SAM). Chapter 3 extends the analysis with an experiment on the significance of age-reading bias. Finally, in Paper IV, we applied a data-moderate assessment method, the stochastic surplus production model in continuous time (SPiCT).

In chapter 4, I express some important perspectives for the assessment and management of these demersal living resources off Namibia. I round off with a conclusion emanating from the new findings.

The Appendixes I – IV are the four papers that were produced during this study.
Chapter 2: Background

In the following, I will briefly describe the Benguela Current Large Marine Ecosystem. Then I describe the biology and ecology, monitoring programs, assessment and management of the four demersal species in the northern Benguela, i.e. the waters off Namibia (Hutchings et al. 2009; Jarre et al. 2015). This is followed by a discussion of three general current trends in fisheries stock assessment, specifically, looking at the newdevelopment of state-of-the-art stock assessment tools and methods. First, the advancement in data-moderate methods, then the more data-demanding approaches that include more biological processes, and finally a discussion of the move to open source models.

2.1 Benguela Current Large Marine Ecosystem

The coastal region of Namibia is within the Benguela Current Large Marine Ecosystem (BCLME) (Figure 1) (Shannon 1985; Hamukuaya et al. 2016). This ecosystem is one of the world’s four major eastern boundary systems characterized by active upwelling (Shannon 1985). This LME is bounded by two warm water regimes: the Agulhas on the south coast (between 36˚ and 37˚ S) and the Angola-Benguela Frontal zone in the northern Namibia/southern Angola (Shannon 1985). The upwelling activities are generated by the intense equator-ward wind stress pattern that exists over the Benguela current (Shannon 1985), resulting in some of the world’s most productive fishing grounds (Crawford et al. 1987).
Figure 1: Conceptual image showing the main physical features and surface current in the Benguela Current Large Marine Ecosystem. Gray shading shows continental shelf, dotted lines are front, the red arrows indicate the warm current and the blue arrows represent the cold current. (Source: BCC, 2008).

The ecosystem is shared between Angola, Namibia and South Africa, a situation that underscores the need for collaboration in the region to ensure sustainable utilization of all transboundary fish stocks. The BCLME is commercially exploited by a range of key economic sectors such as fisheries, petroleum, coastal mining, tourism and marine transportation. The total commercial catch in the Namibian EEZ in 2017 of all (finfish, crustaceans and seals) species under quota management control was over 534 000 tonnes, of which 31% was
hake and 60% was horse mackerel (MFMR 2018). The Cape monkfish fishery is the third highest contributor to the total catch, while west coast sole is the third highest contributor of demersal by-catch.

The region has experienced a number of major environmental changes that had affected the distribution and abundance of commercially important resources (Monteiro and van der Plas 2006). The most dramatic of these was the wide-scale advection of low-oxygen water into the northern Benguela from Angola in 1993 and 1994, and the subsequent Benguela Niño of 1995 (Monteiro and van der Plas 2006), which appears to have severely affected the Namibian sardine population and its major predators (particularly seals), and to have directly or indirectly increased mortality in juvenile hake on the Namibian shelf (Hamukuaya et al. 1998). A recent study by Rouaulta et al. (2018) observed the development and demise of the 2010-2011 Benguela Niño and concluded that it was the warmest event since 1995.

In this thesis, the focus is on hakes, monkfish and west coast sole. These are demersal species with a common habitat and overlapping depth and latitudinal distribution and can be found in the same hauls in what can be referred to as a mixed fishery. Due to their habitat proximity, the removal of the directed fishery has a direct impact on the untargeted species, hence proper assessment should be instituted to avoid depletion of the untargeted resources.

2.2 Biology and ecology of the hake, monkfish and west coast sole

2.2.1 Stock structure

2.2.1.1 Cape hake

Morphologically, the two hake species; *Merluccius capensis* and *M. paradoxus* are very similar (Figure 2 a and b, respectively) in appearance e.g. shape, structure and colour (Gordoa and Duarte 1991; Lloris et al. 2005), but differ in their number of vertebrae (von der Heyden et al. 2007), pigmentation on the gill rakers (Bentz 1976), colour of the anal fin, otolith structure and pectoral fin shape (Gordoa et al. 1995) and genetically (Henriques et al. 2016).
2.2.1.2 Cape monkfish

There are two Cape monkfish species off the Namibian coast (*Lophius vomerinus* and *L. vaillanti*) (Figure 3 a and b, respectively). *L. vomerinus* is the most abundant and 99% of the commercial landings are made up of this species (Maartens 1999).
The population structure of *Lophius* species is poorly known. Analyses of allele frequencies revealed no genetic differentiation between the *L. vomerinus* stock off Namibia and South Africa (Leslie and Grant 1990). A study by Leslie and Grant (1990), found a high level of genetic variation in *L. vomerinus* off South Africa.

Generally, there is limited genetic structure, and low genetic variation in the North Atlantic species (Crozier1989), and the species along the US east coast (Chikarmane et al. 2000).

### 2.2.1.3 West Coast sole

The population structure of west coast sole (*Austroglossus microlepis*) has not been studied, hence, it is poorly known. The shape of *A. microlepis* can readily be distinguished from the other species found in the area by its general body shape (Figure 4), relatively large size and differences in fin ray counts (Smith 1970).
2.2.2 Distribution and migration

2.2.2.1 Cape hake species

The geographical distribution of *M. capensis* and *M. paradoxus* covers the area from 12°S and 14°E in Angola to about 35°S and 30°E off the east coast of South Africa (BCC 2012). These species are therefore among the shared demersal stocks in the BCLME (BCC 2012). *M. capensis* has a higher concentration north of 27°S while *M. paradoxus* is more concentrated south of 27°S (Botha 1985; Burmeister 2001; Johnsen and Kathena 2012). Horizontal movements are observed for both species throughout life, shifting gradually with age indicating substantial alongshore migration (Jansen et al. 2016; Jansen et al. 2017). In the Northern Benguela, the distribution and migration patterns differ between the two species (Jansen et al. 2016; Jansen et al. 2017). *M. capensis* juveniles appeared to move northward away from Walvis Bay, their most likely origin. As they got older, larger and started spawning they progressively aggregate in the vicinity of Walvis Bay, the most important nursery area (Jansen et al. 2016). Considerable quantities of *M. paradoxus* appeared to move northward from the area around Orange River (Jansen et al. 2017).

The vertical distributions also differ between the two species. Both species generally move deeper as they grow older (Gordoia and Duarte 1991; Burmeister 2001). *M. capensis* appeared to stay at depths between a 100-200 m when they are 0.4 to 1.4 years of age, after which they gradually migrate into deeper water (300 m bottom depth) (Jansen et al. 2016). *M. paradoxus* initially appeared to move to shallow water at around 80 m depth, and when they are 0.5 to 5 years old, they gradually move deeper and inhabit waters from 150 to 1000 m bottom depth (Jansen et al. 2017). South of 25°S the distribution of *M. paradoxus* extends to areas shallower
than 200 m depth (Johnsen and Kathena 2012). The two species mainly overlaps in areas with 350-400 m bottom depth (Macpherson and Daurte 1991).

2.2.2.2 Cape monkfish

The distribution of *Lophius vomerinus* extends from northern Namibia (21°S) in the southeast Atlantic to Durban, Natal (30°S, 31°E) on the east coast of South Africa and in the northern and western Indian Ocean (FAO, 2008). *Lophius vaillanti* is found in the eastern Atlantic from north of Walvis Bay (23°S) to the Gulf of Guinea (Maartens and Booth 2005). The bathydemersal *Lophius vomerinus* is found on the continental shelf and upper slope at 150-800 m bottom depths, on soft muddy to hard sand and gravel substrate (Bianchi et al. 1999). The early life stages (egg and larvae) are pelagic (Bianchi et al. 1999).

2.2.2.3 West Coast sole

*Austroglossus microlepis* is distributed from northern Namibia to False Bay (34°S 19°E) in South Africa (Bianchi et al. 1999; Diaz de Astarloa 2002). In the southeast Atlantic, there are three geographically distinct stocks of west coast sole (Payne and Badenhorst 1989). Commercially exploitable quantities of an Angolan stock lie mainly off the mount of the Zaire River, of a northern stock between 20-25°S, south-west and north-west of Walvis Bay and of a southern stock between 28-30°S off the mouth of the Orange River (Macpheson and Gordoa 1992). The northern stock is found in depths ranging from 75-300m, whereas the southern stock has only been fished in depths of 50-100m (Macpheson and Gordoa 1992). The early life stages (egg and larvae) are pelagic while adults live close to the seabed (Bianchi et al. 1999).

2.2.3 Age and growth

2.2.3.1 Cape hake species

*Magalorhinos capensis* and *M. paradoxus* attain a maximum size of 149 and 127 cm total length and a maximum weight of 6 and 4 kg respectively (Paulus et al. 2013). The age of *M. capensis* and *M. paradoxus* are determined from annual rings formation on otoliths that can be counted with a microscope (Wilhelm 2012). Age determination criteria have been reviewed during several workshops in the region (e.g. BENEFIT 2005), and routine age reading in the Ministry of Fisheries and Marine Resources (MFMR) laboratory. Growth curve parameters of
*M. capensis* and *M. paradoxus* are estimated from otoliths data collected on annual hake scientific surveys (Wilhelm 2012). The otoliths are read annually to determine the age of the fish and to develop an age-length key (Wilhelm 2012). These data are currently used in annual hake stock assessments in Namibia (Kirchner et al. 2012). *M. capensis* generally grow faster and mature younger than *M. paradoxus* (Chlapowski 1974; Botha 1986). Females usually grow faster than males for both species (Macpherson 1976). *M. paradoxus* usually show a higher weight-at-length and lower proportion maturity at length than *M. capensis* (Wilhelm 2012).

A recent validation study on the annual occurrence of otoliths growth zones using seal scat samples shows that *M. capensis* grow at about 12 cm year\(^{-1}\) (Wilhelm et al. 2013). The results of this study suggest that the current method for age determination overestimates age and longevity and under-estimate weight-at-age, length-at-age and maturity-at-age (Wilhelm 2012). The result of this study is not implemented in the current stock assessment of hake.

### 2.2.3.2 Cape monkfish

*L. vomerinus* attains a maximum size of 117 cm total length and a maximum weight of 12 kg (Maartens et al. 1999). Different hard parts (otoliths and illicia) have been used to obtain age estimates for *L. vomerinus* (Maartens et al. 1999). Considerable difficulties were experienced in the interpretation of growth zones in *Lophius* species, especially in older fish (Maartens et al. 1999; Griffiths and Hecht 1986; Duarte et al. 1997). Maartens et al. (1999) experience difficulties using sectioned otoliths. Sectioned illicia were, therefore, considered to be the most suitable hard structure to age *L. vomerinus* (Maartens et al. 1999). *L. vomerinus* is a relatively slow-growing and long-lived species with a lifespan in excess of 10 years (Maartens et al. 1999).

Age determination criteria for *L. vomerinus* have been reviewed during several workshops through the Namibia-Spanish bilateral agreement (Landa et al. 2013). The method was validated using length frequency data, through cohort progression analysis. This is the method currently used in ageing *L. vomerinus* in Namibia and adapted to acquire routine annual age data for current use in stock assessments (Landa et al. 2013).
2.2.3.3 West Coast sole

*A. microlepis* attains a maximum size of 75 cm total length and a maximum weight of 4 kg (Lucks et al. 1973). Currently, there is no age information on *A. microlepis*.

2.2.4 Reproductive biology and recruitments

2.2.4.1 Cape hake species

*M. capensis* mature earlier than *M. paradoxus*, 50% have maturated at the age of 1.34 years and 2.8 years respectively (Paulus et al. 2013). A recent study by Jansen et al. (2015) calculated the Gonadosomatic index (GSI) and found the *M. capensis* GSI to be related to fish length for both females and male in Namibia and South Africa. For *M. paradoxus* the effect of length is not significant for GSI in females in Namibia and males in South Africa.

*M. capensis* and *M. paradoxus* spawn at different locations, depths and times (Jansen et al. 2015). Historically, *M. capensis* is known to spawn in the northern Benguela in two different areas in the 1970s and 1980s: Northern (20–21º30'S) and Central (22ºS–25ºS) (O’Toole 1976, 1978; Assorov and Berenbeim 1983; Olivar et al. 1988). In the 1990s and recently it appears that the spawning areas have shifted southward: Central (22–24ºS) and South (26–28ºS) (Kainge et al. 2007; Jansen et al. 2015; Wilhelm et al. 2015). *M. capensis* spawners were found in shallower water (Kainge et al. 2007; Jansen et al. 2015) whereas *M. paradoxus* spawn offshore at depths of 200-650m (Jansen et al. 2015). *M. paradoxus* are known to spawn off the coast of South Africa and the juvenile and adult fish migrate to Namibia (von der Heyden et al. 2007). In a recent study, Jansen et al. (2015) documented a few *M. paradoxus* off southern Namibia (South of 25ºS) with very large gonads indicating spawning activity in Namibian waters.

Historically in the northern Benguela, peak spawning for *M. capensis* occurred in the shallowest waters in spring, from September to November (O’Toole 1978; Olivar et al. 1988; Sundby et al. 2001). However, recent studies suggest that the main spawning season is the austral winter (July–September, peaking in August) (Jansen et al. 2015), or winter and autumn (Wilhelm et al. 2015). A few spawning *M. paradoxus* were observed in August (Jansen et al. 2015). In the southern Benguela, off the South African South Coast peak spawning for
M. capensis is observed in summer (around January). M. paradoxus spawn along the coast of South Africa with increased spawning intensity observed around March and August – October (Jansen et al. 2015).

A study by Voges et al. (2002) shows that optimal conditions for hake recruitment with values above the long-term average were associated with moderate upwelling strength. Lower and higher upwelling intensities reduced recruitment success to below average (Voges et al. 2002). Similar findings were observed by Kainge et al. (2013). Kainge et al. (2013) also found that spawning stock biomass only affects Cape hake recruitment, significantly when it is combined with an index of upwelling strength during the peak spawning period. The model with the explanatory variables: spawning stock biomass and index of upwelling strength explained 88% of the Cape hake recruitment variability.

2.2.4.2 Cape monkfish

*L. vomerinus* spawn flat gelatinous egg masses, called veils, into the water, which float just below the surface (Leslie and Grant 1990). Larval *L. vomerinus* 15-20 mm long have been taken in pelagic fine-mesh nets, and it is therefore estimated that the eggs and larvae remain in the pelagic zone for 4-8 weeks (Leslie and Grant 1990; Armstrong et al. 1992). During this time, surface currents can transport egg masses and larvae between 400 -800 km (Shannon 1985). The gonad mass of a mature female in spawning condition forms up to 35-50% of the body mass (Walmsley et al. 2005), representing a considerable energetic contribution to reproduction. Female *Lophius* mature at a larger size than male and spawning seasonality varies between species and geographic area. The spawning season of *L. vomerinus* is austral spring (Maartens and Booth 2005). Lengths-at-first maturity are 58.2 and 39.9 cm for females and males respectively (Maartens and Booth 2005). The common reproductive strategy of *Lophius*, releasing eggs in single veils, may facilitate their dispersion and that of the larvae over a great distance, which allows for their protection against predators (Armstrong et al. 1992).

Historically, the International Commission of the South East Atlantic Fisheries (ICSEAF), and in particular Spanish researchers, identified two separate recruitment areas; the first being off Walvis Bay (23ºS - 25ºS) at depths between 150 and 300m and the second near the Orange River 28ºS at depths between 100 and 300 m.
(ICSEAF 1984). These observations were confirmed by independent data collected by the Norwegian RV Dr Fridtjof Nansen during a trawl survey in 1990. A similar observation was recently confirmed by Erasmus (2018) through an *L. vomerinus* port sampling program. Erasmus (2018) has also confirmed the previous finding that peak spawning for *L. vomerinus* in northern Benguela occur between July and September, similar to earlier finds by O’Toole (1978) and Maartens and Booth (2005).

2.2.4.3 West Coast sole
Historically, the coastal water between Cape Cross (21º30'S) and Hollam’s Bird Island (24º30'S) is an important spawning location for *A. microlepis* (O’Toole 1978). The abundance and length composition of larvae indicates a relatively short spawning season from early spring until early summer, rapidly decreasing thereafter (O’Toole 1978). Spawning was generally confined to upwelling areas within 30 km from shore and is apparently heaviest in the vicinity of Conception Bay (24ºS) and Hollam’s Bird Island (24º30'S), and between Cape Cross (21º30'S) and Walvis Bay (23ºS) (O’Toole 1978).

2.2.5 Feeding ecology

2.2.5.1 Cape hake species
Both *M. capensis* and *M. paradoxus* are opportunistic feeders that consume available prey, including smaller conspecifics (cannibalism) in relation to the prey’s abundance (Roel and Macpherson 1988; Macpherson and Gordoa 1994). Being opportunistic feeders their diet changes seasonally (Roel and Macpherson 1988). The most common species groups in the hake diets are Krill, crustaceans, cephalopods and other demersal and pelagic fish species (Macpherson and Roel 1987; Traut 1996). The feeding ecology of *M. capensis* and *M. paradoxus* has been studied mainly using stomach content analysis (Macpherson and Roel 1987; Traut 1996). The stomach content results show that *M. capensis* and *M. paradoxus* have a similar diet. *M. capensis* feed actively during the day (07:00 – 13:00), while *M. paradoxus* feed mainly at night (01:00 – 07:00) (Gordoa and Macpherson 1991). A recent study using stable isotopes observed that the values of C13 and N15 were indistinguishable in smaller hake (<30 cm), of both species, suggesting these size classes shared some degree of trophic overlap. In contrast, larger sizes (>30 cm) expressed greater differentiation in both values (Iitembu
et al. 2012). The trophic results also showed that large *M. paradoxus* feed at a higher trophic level than large *M. capensis* (Itembu et al. 2012).

Other than predation by hake, hake is eaten by many organisms such as monkfish (Gordoa and Macpherson 1991), sharks, seabirds, whales, dolphins, fur seals, snoek and other large pelagic predatory fish (Heymans et al. 2004; Roux and Shannon 2004; Mecenero et al. 2006). Estimation of the proportion of the total hake mortality caused by each of the predators is currently not possible due to lack of data. Payne et al. (1987) suggested that hake is the dominant species in its habitat; hence the major predator on hake is probably hake.

### 2.2.5.2 Cape monkfish

Cape monkfish are opportunistic, non-selective feeders, with a common feeding strategy (Gordoa and Macpherson 1990). They are sit-and-wait predators, by camouflaging themselves in the soft sediments and using flicking motions of their illicium to attract forage fishes (Walmsley et al. 2005). Gordoa and Macpherson (1990) suggested that the diet of *L. vomerinus* could be dependent on the reaction of prey species to the lure of the illicium. Prey species such as shallow-water Cape hake *Merluccius capensis*, Gobies *Sufflogobius bibarbatus* and Flying squid *Todarodes sagittatus*, that themselves feed chiefly on mobile prey like euphausiids, decapods, and other mobile prey items, were caught more readily by *L. vomerinus* (Gordoa and Macpherson 1990). Macpherson (1985) noted that *L. vomerinus* was a daytime predator remaining inactive once it has captured its prey, and that food-intake was lowest during January-February and highest in June-August. A recent study by Erasmus (2018) confirmed previously known *L. vomerinus* forage species such as Jacopever *Helicolenus dactylopterus*, shallow-water Cape hake *Merluccius paradoxus*, Gobies *Sufflogobius bibarbatus*, Shortnose greeneye *Chlorophthalmus agazizi*, West Coast sole *Austroglossus microlepis* etc. Information on food habitats and diet composition of *L. vaillanti* is limited. Very low incidences of cannibalism have been reported for *L. vomerinus* (Walmsley et al. 2005; Erasmus 2018). The diet spectrum is size-dependent; small juveniles mainly feed on crustaceans and cephalopods, while large juveniles and adults feed on a wide range of pelagic and benthic fish prey (Gordoa and Macpherson 1990). The large *L. vomerinus* are thus among the top demersal predators in the BCLME. *L. vomerinus* are eaten by various predacious fish including swordfish and sharks (Ebert 1994).
2.2.5.3 West Coast sole
Adult west coast sole prey on worms, crustaceans, molluscs and fish such as gobies. *A. microlepis* constitute a large part of the diet of monkfish (Walmsley et al. 2005; Erasmus 2018). Diet composition by percentage frequency of occurrence and index of relative importance for *L. vomerinus* shows that *A. microlepis* is an important prey (Walmsley et al. 2005).

2.3 Main demersal fishery in Namibia

2.3.1 Cape hake species
The hake fishery off Namibia for *M. capensis* and *M. paradoxus* is the main demersal fishery in the northern Benguela (Burmeister 2001). This fishery is the most important commercial resource in terms of revenue, export value and employment (Wilhelm et al. 2015). These species can be found in the same hauls. This fishery can, therefore, be referred to as a mixed fishery.

The Namibian hake fishery has a relatively long exploitation history comprising of three different management periods (Figure 5): 1964-1976 (open access), 1977-1989 (International Commission for the Southeast Atlantic Fisheries (ICSEAF)) and from 1990 to date (post Namibian independence). During the open access period there were no management measures in place. The annual landings increased sharply during the 60’s and peaked at more than 800 000 tonnes in 1973. After Namibia gained independence in 1990, the Namibian Ministry of Fisheries and Marine Resources (MFMR) took over the management of the resources. The overall management strategy was changed, TACs were set very low and increased gradually, but still remained relatively low when compared to the ICSEAF period (Figure 5).
Figure 5: Namibian hake (Merluccius capensis and M. paradoxus combined) stocks landings data from 1964 to 2017 under three different management period (open access, ICSEAF and post-independence), and the allocated total allowable catches from 1977 to 2017.

Currently, the hake TAC fluctuates between 130 000 and 154 000 tonnes, while the fishing fleets declined from 150–84 vessels (2010–2017). The hake sub-sector is the major source of employment in the fishery sector employing about 10 000 of the total of 15 000 people in 2017 (MFMR 2018). The hake stocks in Namibia are vital for earning of foreign currency to ensure economic success. Hence the resource should be properly assessed and managed sustainably.

2.3.2 Cape monkfish

*L. vomerinus* supports an important commercial fishery in Namibia and is the third most economically valuable demersal fishery in Namibia (Maartens and Booth 2001, 2005) and in South Africa (Walmsley et al. 2005). The Cape monkfish is now a resource of major commercial importance, for which 99% of the commercial landings are *L. vomerinus*, and 1% *L. vaillant* (Maartens 1999). *L. vomerinus* was landed as by-catch in the hake directed fishery up until 1994 (Maartens 1999). Currently, *L. vomerinus* is a targeted fishery managed through a TAC regulation (Figure 6). Due to habitat proximity of Cape hakes and *L. vomerinus*, approximately 5-7% of the *L. vomerinus* TACs, is allocated to the hake directed fishery as unavoidable by-catch.
Landing statistics for monkfish date back to 1974 (Figure 6). Landings recorded by the International Commission for the Southeast Atlantic Fisheries (ICSEAF) between 1974 and 1989 increased to a peak in excess of 16 000 tonnes in 1982. Since Namibian independence in 1990, the annual monkfish landings have been increasing to a peak of around 17 000 tonnes in 1998 and decreased thereafter (Figure 6).

![Figure 6: Namibian monkfish (Lophius vomerinus) landings statistics from 1974 to 2017 and the allocated total allowable catches from 2001 to 2017.](image)

Historically, *L. vomerinus* was the main by-catch in the hake fishery; however, due to the escalation of its economic value, it has resulted in the development of a fishery targeting *L. vomerinus* (Maartens and Booth 2005). In 1994, a monkfish target fishery was established, managed through effort control, with limited access and a restriction of 800 horsepower on vessel capacity (Maartens et al. 1999). Currently, the monkfish TAC is set around 9 000 – 13 000 tones, harvested by 19 - 24 fishing vessels targeting *L. vomerinus*.

### 2.3.3 West Coast sole

*A. microlepis* was important commercially since 1960, but was heavily overfished (1969-1971). The landings of *A. microlepis* peaked at nearly 2000 t in 1972, but have declined steeply since then, currently amounting to
less than 800 tonnes per year (Lucks et al. 1973; Diaz de Astarloa 2002). In the northern Benguela, *A. microlepis* is commercially the only important flatfish. *A. microlepis* is generally captured as by-catch during bottom trawling for demersal species such as Cape hakes and Cape monkfish (Bianchi et al. 1999). Economically, *A. microlepis* is the most important by-catch species in the Namibian Cape monkfish fishery. *A. microlepis* landings represent a relatively small fraction of the total demersal landings in the northern Benguela. Landings statistics for *A. microlepis* from the monkfish directed fishery between 1999 – 2017 are shown in Figure 7. The annual landings increased to a peak of around 900 tonnes in 2007 and decreased thereafter (Figure 7).

Since 1970, the *A. microlepis* landings have predominantly come from the area between Rocky Point (19°S) and Palgrave Point (20°30’ S) (O’Toole 1977). This was particularly evident during the peak *A. microlepis* years 1971 to 1974. Only a small portion of *A. microlepis* landings resulted from fishing in more extensive grounds south of Walvis Bay (O’Toole 1977).

Figure 7: Namibian west coast sole (*Austroglossus microlepis*) landings as by-catch in the monkfish fishery from 1999 to 2017.
2.4 Monitoring

In Namibia, seven scientific surveys are conducted annually, applying methods that include swept-area bottom trawl, mid-water trawl, acoustics, traps and pots. Each survey is directed to one of the following species: hake, monkfish, orange roughy, deep sea red crab, horse mackerel, sardine and rock lobster. These surveys are conducted at different times of the year. All species caught in each trawl are identified to the lowest taxa and sampled, either as total or sub-sample counted, weighed and recorded to the nearest centimetre and gram. Additional biological measures are taken from representative subsamples. This includes sex, weight, maturity and otoliths for age determination. The survey information is added to the Nan-SIS database (Strømme 1992).

By law, during commercial fishing operations, each vessel is obliged to carry a fisheries observer, and a logbook (MFMR 2000). Fisheries observers are required to sample biological information such as length, sex, maturity. Information such as trawl position, date, time, depth, and other vessel details is captured in the daily logbook. Commercial logbooks and biological information from fisheries observers are stored in the Fisheries Information and Management Systems (FIMS).

2.4.1 Cape hake survey

The scientific swept-area surveys targeting *M. capensis* and *M. paradoxus* started in 1990. These surveys were conducted by the Namibian Ministry of Fisheries and Marine Resources onboard the Norwegian R/V Dr Fridtjof Nansen up until 1999 (Sætersdal et al. 1999). The first six years 2–3 surveys per year were conducted and from 1997 to date only one survey per year in austral summer (January–February) (Burmeister 2001). The survey has a target to cover about 200 fixed trawl stations each year.

Figure 8 shows normalized biomass ($\tilde{B}_t = \frac{B_t}{\bar{B}}$) where $\bar{B}$ is the overall mean biomass and subscripts $t$ represents the year. We plotted normalized biomass because the swept-area survey biomass estimates only give an index, and here the reference is the average. There is substantial year-to-year variation within and between *M.*
capensis and *M. paradoxus* (Figure 8). Further investigations and data will be required to conclude on this recent apparent strong negative correlation between the survey results of the two hake species.

Figure 8: Normalized biomasses of *M. capensis* (dashed black line) and *M. paradoxus* (dashed red line) with respective standard deviations shown. These indices are shown since the establishment of the hake biomass swept-area survey by the Ministry of Fisheries and Marine Resources in 1990. Source: Data extracted from the Nan-SIS database (Strømme, 1992), a survey database for the Ministry of Fisheries and Marine Resources.

### 2.4.2 Cape monkfish survey

Cape monkfish are monitored through an annual swept-area scientific survey. The surveys started in 2000 onboard the R/V Welwitchia up until 2014, and from 2015 onward surveys were conducted by R/V Mirabilis. The survey has a target to cover about 94 fixed trawl stations each year and is conducted during the month on November. Figure 9 shows a normalized biomass \( \bar{B}_t = \frac{B_t}{\bar{B}} \) where \( \bar{B} \) is the overall mean biomass and subscripts \( t \) represents the year. Normalized biomass results show a declining trend since the inception of the monkfish scientific survey in 2000 until 2005. In 2006, the survey was not conducted. The normalized biomass was above average between 2009-2011, and has been fluctuating above and below average until 2016 (Figure 9).
2.5 Assessment of the main demersal stocks in Namibia

In Namibia, stock assessments are the basis for advice to the fisheries managers. Mathematical models and data from surveys and commercial fisheries are used to conduct the assessment. Hence, there is a high reliance on statistical analysis, when providing management advice.

2.5.1 Cape hake assessment

At present, *M. capensis* and *M. paradoxus* are assessed and managed as a combined single stock in Namibia (Kirchner et al. 2012; Kathena et al. 2016). Management advice and TAC allocation are based on a statistical catch-at-age analysis (SCAA) model fitted to data (Kirchner et al. 2012; Radermeyer et al. 2008). This assessment ignores species-specific dynamics. SCAA uses estimates of weight-at-age, maturity-at-age, natural mortality and survey biomass coefficient of variation (CV) derived from *M. capensis*, thus ignoring the dynamics of *M. paradoxus*. The current assessment is comprised of twelve different model specifications: a base case and eleven sensitivity tests (Kirchner et al. 2012). The sensitivity tests are designed to consider different scenarios, such as assuming age independent natural mortality, age-dependent natural mortality,
omitting some of the commercial information, excluding historical surveys, etc. (Kirchner et al. 2012; Wilhelm et al. 2015). The selection of the best fit model is based on the Akaike Information Criteria.

2.5.2 Cape monkfish assessment
Currently, *L. vomerinus* stock is assessed annually using an Age-Structured Production Model (ASPM, Rademeyer and Nishida 2011). Input data to the model are stock and catch weight-at-age, maturity-at-age, natural mortality, survey biomass index, commercial catch rates, survey catch-at-age and commercial-catch-at-age. The current assessment of *L. vomerinus* has indicated that the spawning stock biomass is above the spawning stock biomass leading to the maximum sustainable yield (MSY) level.

ASPM approach involves constructing an age-structured model to simulate the population dynamics of the stock and tuning it to available indices by maximizing a likelihood function. The number of recruits at the start of each year is related to the previous year’s spawner stock biomass using the Beverton and Holt stock-recruitment relationship.

2.5.3 West Coast sole assessment
*A. microlepis* has never formally been assessed and there are no conservation measures in place.

2.6 Management of the main demersal stocks in Namibia
Annually, for each fishery, the ministry scientists present the TAC recommendations to the Marine Resource Advisory Council (MRAC). MRAC consists of members from the fishing industry, banking sector, labour union, legal fraternity and fishery experts. MRAC considers the scientific advice and drafts recommendations to the Minister of Fisheries and Marine Resources. The minister then presents these recommendations to the cabinet, and the cabinet decides on the TAC. The cabinet decision is passed on to the Attorney General’s office for gazette (legally-binding publication). Once gazetted, the officials from the ministry of fisheries will allocate quotas to individual fishing right holders.

2.6.1 Cape hake management
The International Commission for Southeast Atlantic Fisheries (ICSEAF) came into force in 1972 (Butterworth and Rademeyer 2005). ICSEAF divided the Namibian coastline into three management divisions:
Division 1.3 (17 – 20˚ S), Division 1.4 (20 -25˚ S) and Division 1.5 (25 – 30˚ S). Management of the hake fishery in the northern Benguela began in 1977, with the allocation of TAC by the ICSEAF (Butterworth and Rademeyer 2005). The fishery over-caught the first set TAC.

After independence in 1990, the Ministry of Fisheries and Marine Resources (MFMR) took over the management of the fishery. At that time the hake stocks were considered to be heavily depleted (Payne and Punt 1995). From 1990 to 1994 the MFMR took several measures to protect the hake stocks, including a ban on foreign fishing fleets, proclamation of a 200-mile-EEZ, according to international law, 200-m depth restriction (no hake-directed fishing allowed shallower than the 200-m-isobath), a minimum mesh size set at 110 mm, aimed at protecting <36 cm hakes, and TAC reduced to 60 000 tonnes (HMP 2014). The government furthermore established institutions and implemented policies to monitor fishing and regulate fishing rights in Namibian waters, such as the introduction of the national policy on exploitation rights and quota allocation, and the establishment of the Directorate of monitoring, control and surveillance. An observer program was also established and all hake fishing vessels are required to carry a Fisheries Observer onboard, as well as offload their catches at either Walvis Bay or Luderitz ports under the inspection of the Fisheries Inspectors. The role of a Fisheries observer onboard is to ensure that the vessel complies with the fishing regulation and to collect scientific information. In 2006, an additional measure was introduced restricting hake fishing shallower than 300 m bottom depth south of 25 ºS (Johnsen and Kathena 2012).

During the years 1990 to 1996 TAC’s were equated to 20% of the fishable biomass estimates from the research swept-area biomass survey. This was assumed to be a conservative approach (Butterworth and Rademeyer 2005; Butterworth and Geromont 2001). Fishable biomass in this context was defined as the biomass of fish that have attained a total length of > 36 cm, at the time of a survey. The survey estimates during that period fluctuated and created great uncertainty about the status of the hake resource. The reliability of this method in setting TAC was heavily debated. These debates resulted in the development of the Interim Management Procedures (IMP) in 1997. The IMP was structured such that it could adjust the recommended TAC up or down depending on trends in the research survey and the commercial CPUE data (Butterworth and Geromont 2001). The purpose of the IMP’s was to provide a recommended TAC in a precautionary way and to resolve
the debate on the survey estimates, whether survey estimates should be treated as absolute or relative abundance indices in stock assessment models (Butterworth and Geromont 2001). The IMP was used for management from 1998 to 2001 and confirmed that survey estimates should be treated as relative abundance indices in stock assessment models (Butterworth and Geromont 2001). In 2002, an Operational Management Procedure (OMP) was implemented and served for the subsequent three years (HMP 2014). In 2005, an SCAA model was developed. Since 2005, advice and subsequent management has been based on the output from this model, (Kirchner et al. 2012).

2.6.2 Cape monkfish management
In 1994, fishing rights to catch *L. vomerinus* with a hake by-catch quota were implemented. The monkfish fishery was managed through effort control, with limited access and a restriction of 800 horsepower on vessel capacity until the year 2000 (Maartens 1999). From 2001 to date, the monkfish is a targeted fishery managed through a TAC regulation (Maartens 1999). Due to habitat proximity of these stocks a certain percentage, approximately 5-7% of the Cape monkfish TAC, is allocated to the hake directed fishery as unavoidable by-catch. For the past decade the TAC for *L. vomerinus* has been fluctuating around 10 000 tons.

2.6.3 West Coast sole management
The fishery is managed by setting by-catch levies. These are set at a high level, to deter operators from deliberately targeting this species.

2.7 Three recent trends in fisheries stock assessment
Three particular recent trends in fish stock assessment have had a major impact on designing and implementing this thesis to produce a promising platform for improving the assessment and management of Namibian demersal fish stocks. First, the increasing expectation by decision-makers in jurisdictions such as Australia, New Zealand, and the United States (Punt et al. 2013), and non-governmental organizations such as Marine Stewardship Council (MSC), that there should be quantitative stock assessments for most if not all exploited fish. Similar interest has recently been expressed by the Namibian government, the Namibian fishing industry, and other stakeholders that all exploited stocks should be assessed to provide scientific advice on optimal harvesting.
Demands for science-based management decisions have spurred a whole new field of data-limited methods (e.g., Carruthers et al. 2014). Alone in the USA, the requirement for annual catch limits (ACLs) has spurred the development and adoption of at least 16 methods for establishing ACLs for data-limited fisheries (Newman et al. 2015). Length-based assessment methods constitute an important part of data-limited methods because it is much easier to obtain reliable length measurements of catch samples than to measure total catch or record effort data for small-scale or non-target species (Harley et al. 2001; Kokkalis et al. 2015; Rudd and Thorson 2018). It is surprising that with a general understanding of biological parameters it is actually possible to estimate the status of a stock (F/Fmsy) from a length distribution of the catch in just one year (Kokkalis et al. 2017). However, it is impossible to determine the cause of many small fish in the catch from a single year of data with methods based on equilibrium assumptions. Is the cause due to strong year-classes in recent years or to the removal of larger fish from the system? (Rudd and Thorson 2018). For such reasons and because situations exist in Namibia where neither length or age data are available, we focused on surplus production modelling requiring only time-series of commercial catch and biomass or catch-per-unit-of-effort data (Punt 2003; Froese et al. 2017; Pedersen and Berg 2017). We chose the method introduced by Pedersen and Berg (2017), i.e., the stochastic surplus production model in continuous time (SPiCT) to introduce data-limited stock assessments for demersal fish stocks in Namibia.

The perspective in the second trend is nicely formulated by (Thorson 2018): “Let’s simplify stock assessment by replacing tuning algorithms with statistics”. Inspired by Cotter et al. (2004) it is instructive to see this as a broader trend in a longer perspective starting by going a generation back when the development of assessment models was very limited by computational power. Then it was natural to focus only on what could be computed and how. Statisticians had computational tools for linear models and then for general linear models (GLMs) but these tools did not allow much creativity in formulating a model that could explain fisheries data. Instead, models were based on ideas like the Virtual Population Analysis (VPA); *ad hoc* tunings, the extended survivor’s analysis (XSA) and tracking cohorts. Then at some point, it became possible to fit parameters in a model by writing a likelihood function (or a similar measure of fit) and maximizing this numerically with respect to parameters. ADMB (without the Random Effect module) was a very effective tool for this. This
meant more flexibility in describing the stock and the fisheries, and in fitting models to the data. A problem with these models is that they contain a large number of parameters, for example, fishing mortalities. We can’t estimate more parameters than we have data points, so some “prior” is needed to reduce the number of degrees of freedom. A natural way to do this is mixed-effect models, where the fishing mortalities are considered random variables and the model specifies their joint distribution. In a time series context, it is natural to formulate these mixed-effect models as state-space models. Here, it is important to realize that such mixed-effect models were not computationally tractable until recently. The arrival of ADMB-RE changed this and after that Template Model Builder (TMB) (Kristensen et al. 2016). It is now possible to write the model as the biologists see the world, including unobserved random quantities, and then let the software integrate unobserved random variable out and maximize over the parameters. The Thorson (2018) advocacy is in line with the new development in the state-space model framework. State-space models are regarded as powerful tools for modelling time-varying selectivity (Nielsen and Berg 2014) because they simultaneously account for both process and observation errors.

Although SCAA is currently used to assess the combined hake resources off Namibia, we chose to use the new state-space approach SAM (Nielsen and Berg 2014) in this study to represent the new trend for data-rich stock assessment. A preliminary comparison of SAM and SCAA was carried out at the start of this PhD study (Kathena et al. 2015). The models share many of the same assumptions and basic equations but they differ in two important areas. First, SCAA considers observation noise only whereas SAM splits into process noise and observation noise. Second, the hake SCAA model uses a parametric function to describe the selectivity (relative fishing mortality of the different age groups), whereas SAM uses the random walk approach mentioned above as an unobserved stochastic process. The main driver behind the differences observed seems to be linked to the differences between the estimated selectivity (Kathena et al. 2015). SAM is free to estimate the selectivity of any shape, whereas SCAA is constrained by a parametric shape. This SCAA assumption of time-invariant selectivity considered constant in three-time blocks may explain some of the differences since SAM selectivity estimates are time-varying.
The third trend in fisheries stock assessment includes more transparency in model coding and model formulation. A decade or two ago, the analyst would show up at stock assessment workshops with their own model, implemented in excel or in their own codes. The analyst did not share code and often only the author had a full understanding of the model. The new Open-Source trend uses tools such as GitHub (github.org) providing a platform for the model developer to share their models and improve assessment procedures. Such a globally accessible repository provides increased transparency and promotes good practice. This implies that biologists in the future can be more directly engaged in the model building because they can share available models so that all stakeholders can reproduce the results and improve on the models.
Chapter 3: Contributions to the Namibian stock management, fisheries management and stock assessment

This study is aimed at contributing towards the current understanding with regards to stock assessment of demersal fish stocks off Namibia. The contribution should lead to better-informed management advice. This thesis developed an algorithm to estimate the hake species ratio at each commercial haul. This is achieved through spatiotemporal models (Paper I), which utilize the species separated data from the observer programme and/or the scientific surveys. We feed these models with logsheet information on depth and latitude to predict the species ratio of hake species in the individual commercial trawl catches.

Using Paper I as input the thesis proceeds to estimate a species-specific abundance measure for hake at each commercial haul (Paper II). We develop a generalized additive model (GAM) to identify factors that are affecting the separated catch rates of *M. capensis* and *M. paradoxus* in the commercial fishing, and conclude on hake species-specific studies by implementing a data-rich state-space assessment model (Paper III) for two independent single-species assessment, one for *M. capensis*, and the other for *M. paradoxus*.

Finally, we applied the Stochastic Surplus Production model in Continuous Time (SPiCT) (Paper IV), a data-moderate method, to conduct an assessment for west coast sole for the first time.

3.1 Splitting commercial catches of mixed hake into species (Paper I)

Commercial catch data are provided by the fishing industry through the mandatory logbooks. The fishermen log the catch in each catch operation and the entire commercial logsheet database provides the scientists with a unique and highly disaggregated dataset for research and stock assessment. However, the fishermen do not distinguish between the two hake species *M. capensis* and *M. paradoxus* and record them by weight as one single hake category. The work in this paper closes this gap between the aggregated hake catches and the species-specific catches required for beginning to make single-species hake assessment for Namibia (Paper II and Paper III).
We developed models that predict the hake species ratio from explanatory variables including latitude and depth. The input to these generalized linear models (GLM’s) was the species proportion by weight in samples from the commercial hauls obtained through the observer sampling programme. The robustness of the splitting method was examined by comparing differences in model output when the input data instead was taken from the scientific surveys. To facilitate these comparisons, uncertainties were taken into account raising survey catches to adjust for size and species-dependent escapement under the fishing line (which does not occur with the commercial trawl, because there is no gap between the fishing line and the ground gear) and then reducing catches by the codend selectivity in the commercial trawl (since codend selectivity in the survey trawl is negligible for commercial hake sizes). Finally, the selected model was used to split catches in the logbook into species, providing a dataset with catch per species in the individual commercial hauls. The process and data flow are illustrated in Figure 10.

This work was fully detailed and published in African Journal of Marine Science in 2012 as: Johnsen, E., Kathena, J., 2012. A robust method for generating separate catch time-series for each of the hake species caught in the Namibian trawl fishery. African Journal of Marine Science, 34: 43–53. This peer-reviewed paper is attached as paper I in the present thesis (see Appendix I).
Figure 10: Flowchart illustrating the steps and data used in predicting the species catch ratio by weight in commercial catches. Note independent models of species ratio by weight with depth, latitude and (year, quarter) were developed from survey data and observer data respectively.

A brief description of the methods and highlights of the results are given below.

3.1.1 Data

Commercial samples were extracted from the Fisheries Information and Management Systems (FIMS) database administered by NatMIRC. These samples were collected by fisheries observers onboard the commercial trawlers. This involved samples from selected hauls for length measurements (200 fish) and/or biological measures (80 fish) such as sex determination and maturity staging. Raising to the total catch was done by the ratio of total weight to estimated sample weight. The commercial trawlers use a minimum codend mesh size of 110 mm, allowing some escapement of small hake, and the majority of the trawlers used trawls with a vertical net opening of 4-8 m (Johnsen and Lilende 2007).

Survey length frequency distribution data were obtained from the NAN-SIS database of demersal biomass surveys conducted by NatMIRC (Strømme 1992). There were 15 surveys in January –February (austral
summer) from 1993 to 2008 and five winter/autumn surveys in April - June and September - November (Burmeister 2001). The survey vessel used a Gisund Super two-panel bottom trawl with 20 mm outer-codend mesh, lined with 10 mm inner-net, with the vertical net opening of 4.2-4.5 m. From each trawl haul, a total catch or a sub-sample of *M. capensis* and *M. paradoxus* was measured for fish total length (TL). Sub-samples were raised to the total catch by weight.

### 3.1.2 Modelling

We made two sets of GLM models, one using the observer data, and another using the survey data. However, to make these models compatible, the data from the survey trawl was preprocessed to simulate comparable data from the commercial trawl. This preprocessing included two steps:

1. We first considered the species-and size-dependent escapement under the fishing line that was demonstrated by Jørgensen et al. (2007) using the Gisund Super trawl. They mounted a collecting bag under the trawl so that all fish escaping under the fishing line was captured in the collecting bag. Assuming that neither *M. capensis* nor *M. paradoxus* escaped above the headline of the trawl, the proportion of fish retained by the trawl at each of the trawl stations was calculated as the number of fish in the codend divided by the total number of fish in the codend and in the collecting bag for each class. Jørgensen et al. (2007) modelled retention (*r*) as a function of fish length (*l*) by a logistic regression with the logit link function

   \[ r(l) = \frac{e^{a+bl}}{1+e^{a+bl}} = \frac{1}{1 + e^{-b(l + a/b)}} \]

   where *a* and *b* are generic parameters. From the bag-experiments, they estimated consistently negative values of *b* for both species implying that retention decreased with increasing fish length, i.e. larger fish escape under the fishing line more frequently than small fish. The parameter *a* was estimated positive giving positive values of *L*50 = -*a/b*, the length at which retention has decreased to ½. We used averaged values of *a* and *b* from their bag-experiments. However, to account for the considerable variations in haul-specific retention curves obtained by Jørgensen et al. (2007), we used stochastic simulation (see input data in Table 2 Paper I) to generate random *a* and *b* parameters by the “mvnorm” function (Venables and Ripley 2004) in R. The retention parameters for *M. capensis* were considered independent of the retention parameters for *M. paradoxus*.

Thus for each survey station, we simulate two new independent retention curves for the two species. Figure 11 shows 20 such curves for *M. capensis* (top graph) and 20 curves for *M. paradoxus* (middle graph). The escapement is considerably smaller (i.e. higher retention) with a narrower distribution for *M. capensis* than the more variable and pronounced escapement increase with fish size for *M. paradoxus*.

This first step was concluded by dividing for each survey station the catch rate by species and length class by a simulated survey selectivity (*r*) to obtain the estimated catch if there was no escapement under the fishing line (Eq. 2 Paper I)
2. The second step addresses the difference in codend selectivity between the two trawls. We use standard S-shaped selection curves (a<0, b>0, Table 2 Paper I). Since no study had been conducted on codend selectivity in the Namibian hake fishery, we extracted the selectivity parameter from a study by Galvez and Rebolledo (2005). Galvez and Rebolledo (2005) studied codend mesh size selectivity of bottom trawl nets in the Chilean hake (Merluccius gaygayi) fishery for four mesh sizes openings in the codend (100, 110, 130 and 140mm mesh openings) and using a covered codend method. Simulations of the applied codend mesh size selectivity for Namibian hake fishery are shown in Figure 11 (bottom) using a 110 mm codend. All the selection runs in Figure 11 take into account the fact that species-specific a and b parameters are not entirely independent (covariance is given in Table 2 Paper I based on Jørgensen et al. (2007) for the retention and Jørgensen et al. (2006) for codend selectivity of cod).

This second step was concluded by multiplying the catch rates with this simulated codend selectivity of hakes in the commercial trawl. Converting catches to weight the proportion of M. capensis in the catch at each survey station could be calculated (spr Eq. 6 Paper I).

Figure 11: Simulated retention curves and simulated mesh size selectivity (n=20). The two top panels express that large fish are able to escape under the fishing line while the bottom panel shows that small fish can escape through the mesh; (top panel) Merluccius capensis, (middle panel) M. paradoxus and (bottom panel) used to simulate codend mesh size selection in the commercial trawl with a 110 mm codend (Figure 1 from Paper I).
We applied GLMs to derive the spatial-temporal patterns of species ratio using temporal (year, quarter) and spatial (latitude, depth) data from the observer and survey datasets. The modelling was done to establish factors affecting the species ratio since the fishery is unevenly spread in time and space across a gradient. Modelling of the species ratio (M. capensis and M. paradoxus) was implemented with species ratio as the response variable. The explanatory variables used are latitude, depth, year and quarter. The models applied in this study are listed in Table 1 and 2. The model fitting was done in R (R core Team, 2012) using the lm (package) (Wood, 2011). Model selection was based on the explanatory power of the different variables and chi²-based tests of the residual deviance and p-value. Model 1 in both tables provided the best fit for both data and was selected for further analysis.

Table 1: Model specification for the survey length species ratio, called spr in Eq. 6 Paper I

<table>
<thead>
<tr>
<th>Model #</th>
<th>Predictors</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>(LSR ~ depth + latitude, family = binomial(link=logit))</td>
</tr>
<tr>
<td>2</td>
<td>(LSR ~ depth + latitude + quarter, family = binomial(link=logit))</td>
</tr>
<tr>
<td>3</td>
<td>(LSR ~ depth + latitude + quarter + year, family = binomial(link=logit))</td>
</tr>
<tr>
<td>4</td>
<td>(LSR ~ depth + latitude + year, family = binomial(link=logit))</td>
</tr>
</tbody>
</table>

Table 2: Model specification for the observer species ratio, called spro in Eq. 7 Paper I

<table>
<thead>
<tr>
<th>Model #</th>
<th>Predictors</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>(SPR ~ depth + latitude, family = binomial(link=logit))</td>
</tr>
<tr>
<td>2</td>
<td>(SPR ~ depth + latitude + quarter, family = binomial(link=logit))</td>
</tr>
<tr>
<td>3</td>
<td>(SPR ~ depth + latitude + quarter + year, family = binomial(link=logit))</td>
</tr>
<tr>
<td>4</td>
<td>(SPR ~ depth + latitude + year, family = binomial(link=logit))</td>
</tr>
</tbody>
</table>
3.1.3 Result highlights

The species catch ratio shows that in the north and central region (17° S to 24° S) there is a higher ratio of *M. capensis* in shallow water (<400 m) and very few in deeper water (>400 m bottom depth) (Figure 12). While in the south region (26° S to 29° S) the ratio of *M. capensis* in shallow water decreases southwards and there is hardly any *M. capensis* in water deeper than 350 m bottom depth (Figure 12).

![Species catch ratio per depth and latitude](image)

Figure 12: Species catch ratio per depth and latitude, based on observer data from 1998-2009, computed from the best model. Greyscale bar corresponding to the *M. capensis* ratio over the depth-latitude distribution.

With the model of species ratios in the catch, and a time series of combined commercial catches of the two species, it is possible to split these catches, i.e. estimate the catches of each species. With latitude and depth as explanatory variables, the predicted proportion of *M. capensis* in the catch by quarter using survey data gives almost the same results as the observer based results throughout 1998-2007 (Figure 7a in Paper I). Furthermore
the effect of including quarter as an explanatory variable appears negligible when based on observer data. Figure 7c (Paper I) also demonstrates that although our simulations result in considerable variation in the *M. capensis* species proportions (spr) with increasing species overlap (spr between 0.3 and 0.7 in Figure 3 Paper I), this variation seems to disappear when calculating the quarterly averages using percentage of *M. capensis* in the individual hake catches by stations as relative weights. Figure 13 shows the estimated yearly commercial catches time series of *M. capensis* and *M. paradoxus* for the period 1998-2014. The catch trends are very similar throughout the studied period. Nearly 70% of the landings from the hake directed fishery was estimated to be *M. paradoxus*.

![Figure 13: Stacked total catch of *M. capensis* and *M. paradoxus* in tonnes from the hake trawl fleets off the Namibian coastline for the period 1998–2014, representing annual catch input to Paper II.](image)

### 3.1.4 Discussion

This analysis of species catch ratio was based on species-separated survey and observer data. Both datasets are subsampled onboard commercial vessels during fishing operations and onboard the research vessel during the scientific surveys. GLM’s were applied to these datasets to determine the factors affecting the species catch ratio. GLM analysis demonstrated the value of considering depth and latitude when estimating the species catch ratio. Our results support previous and more recent findings (Gordoa et al. 1995; Strømme et al. 2016;
Jansen et al. 2015; Henriques et al. 2016) that, *M. capensis* and *M. paradoxus* in the northern Benguela have partially overlapping depth-latitudinal distributions where more than 75% of the *M. Capensis* are found shallower than 300m depth while the same percentage of *M. paradoxus* is found deeper than 300 m depth (Burmeister 2001). Overlapping distributions and the morphological similarities have led to the recording of the two hake species as one single hake category in the commercial landings. Thus, species-specific assessments of the two hake species in the northern Benguela have not been conducted. We show here that species separated data from surveys and observer programme can be used to split the commercial catch into species-specific catches of each commercial haul.

The survey data was preprocessed to account for escapement under the fishing line and the codend mesh size selectivity. Furthermore, both availability to and escapement from the trawl vary between species and with fish size and are also affected by individual behavioural and ecological factors such as changes in the environment, and an individual’s motivation, for example, spawning and feeding. All these factors will certainly affect the relationship between the relative catch of each species (Hjellvik et al. 2003). The use of empirically based and realistically wide-ranging selectivity curves cannot compensate for unknown factors such as the position of fish in the water column and their reaction to the survey vessel and catching equipment. Jørgensen et al. (2007) also concluded that further bag-experiments are required to understand the marked depth and fish size effects on escapements of *M. paradoxus*. Thus, at present, it is still not known what causes the difference between the proportion of *M. paradoxus* as a function of depth in the survey data and in the observer data.

The observer sampled data onboard the commercial fishing fleets have some advantages. First, the seasonal sampling coverage of the observer data is much higher than in the surveys. Second species proportions based on commercial samples are not affected by the uncertainties in the retention modelling as it is for the survey. However, the disadvantages of using observer data are some risk of mis-species identification (cf Figure 4, Paper I) and sampling biased towards larger vessels. But everything taking into account has favoured the observed based species splitting model.
The spatiotemporal variability in the distribution of the two hake species and the developed procedure may lack flexibility during periods of abrupt species displacement, as occurred in 1994 when large amounts of juvenile *M. capensis* migrated offshore to avoid hypoxic water and were subsequently caught by commercial trawlers (Hamukuaya et al. 1998). A model that allows for changes in the spatial pattern in species ratio (spro) over time could be an important extension of the current splitting procedure. Furthermore, Paper II indicates that for example vessel (gear) characteristics also have an effect on spro. Commercial size classes used as a categorical variable should also be pursued. Possibly this also gives rise to significant interactions between variables and the task of improving the species splitting algorithm and examining their consequences on the assessments deserves a PhD study in itself.

In conclusion, the hake species splitting for Namibia is still to be considered a challenge in particular concerning escapement under the fishing line if survey information is used. With increased computational abilities in the future, it should become feasible to implement an integrated approach such as using all data simultaneously in performing splitting by length and GAM CPUE standardization (Paper II and section 3.2.2)

### 3.2 Commercial catch rate analysis (Paper II)

Based on the results from paper 1, commercial catch rates or catch-per-unit-of effort (CPUE) of *M. capensis* and *M. paradoxus* have been calculated for each trawl haul that targeted hake in Namibian waters from 1998-2014. These data are potentially useful as input to stock assessment models. Many assessment models, be they data-limited, data-moderate or data-rich, use CPUE as a relative index of fish stock abundance. However, the annual mean of CPUE by haul cannot be used directly in stock assessment models because the fishery is unevenly spread in space and time across a gradient that affects the CPUE. The CPUE of a single haul is affected by many other factors than the total fish stock abundance, such as the time of the day, the fishing vessel characteristics, and the position. Not accounting for these factors may, therefore, bias the CPUE index and lead to wrong conclusions. The utility of indices of abundance based on catch and effort data can thus be improved by standardizing them; that is, removing from the data any variation due to effects other than fish abundance (Punt et al. 2000). This process is often referred to as catch-effort standardization.
The aim of this study was to standardize the dataset of haul-by-haul species-specific catch and effort data. The study identifies factors (time, position, depth, vessel and light levels) affecting the CPUE of *M. capensis* and *M. paradoxus* so that the effect of these factors can be removed from the CPUE time series. Furthermore, the estimation of these effects provides ecological and behavioural information of these species in northern Benguela.


A brief description of the methods and highlights of the results are given below.

### 3.2.1 Data
We used a unique dataset of haul-by-haul species-specific catch and effort data, which was developed in Paper I, where total hake catches had been split into *M. capensis* and *M. paradoxus* based on the observer data-based model. The size of the database was approximately 764 633 individual trawl hauls for the period 1998-2014. The datasets contained haul-by-haul commercial trawl catches of *M. capensis* and *M. paradoxus* by vessel ID, vessel type (wet trawler or freezer trawler), vessels characteristics (gross registered tonnage, length, and horsepower), trawl position, date, time and depth.

### 3.2.2 Modelling
We used GAMs to model the catch rates (CPUE, in Kg per hour-of-trawling) of species (either *M. capensis* or *M. paradoxus*), with catch-per-species as the response variable and the effort (trawl duration in hours) as an offset variable. We developed nine GAMs, for each species using different combinations of the explanatory variables; time, latitude, depth, light levels and vessel. The model with the lowest AIC value for each species was selected as the best model.
3.2.3 Result highlights

Standardized catch rates for *M. paradoxus* were significantly higher than catch rates of *M. capensis*. The temporal patterns in the time-series of the two species were similar and indicated a series of low values between 2002 and 2007 and a peak in 2011 (Figure 14). Yearly distribution maps also indicated interannual variations (Figure 15). The *M. capensis* distribution (Figure 15 a and b) shows a shift in distribution towards deeper water in 2011, while *M. paradoxus* shows a higher density around 17° S in the year 2014, which was not observed the previous year (Figure 15 c and d).

![Graph showing yearly effect of catch rates of *Merluccius capensis* and *M. paradoxus* in the northern Benguela system, predicted by generalized additive models (GAMs) model #1.](image)

Figure 14: Yearly effect of catch rates of *Merluccius capensis* and *M. paradoxus* in the northern Benguela system, predicted by generalized additive models (GAMs) model #1. The thin dashed lines represent the 95% confidence interval from the GAM and the central line is the predicted mean. The upper limit confidence interval extends to 7600 kg/hr for *M. paradoxus* in 2011 (Figure 3a from paper II).
Figure 15: Yearly difference in spatial distribution of catch rates according to depth and latitude of (a, b) *Merluccius capensis* is 2010 and 2011, and (c, d) *M. paradoxus* in 2013 and 2014, respectively, in the northern Benguela system (Figure 6 from paper II).

### 3.2.4 Discussion

We standardized the dataset of haul-by-haul species-specific catch and effort data and described seasonal patterns. Seasonal migration patterns were found to correspond to the known spawning areas and season for *M. capensis* and *M. paradoxus*. The study also found that yearly standardized CPUE time-series are problematic as proxies for total stock abundance because of spatial coverage issues. Consequently, such CPUE data should not be used for stock-size assessment and fisheries advice concerning northern Benguela hake until this is solved.
The present analysis of catch rates was based on the data set produced by applying the method developed in the paper I, i.e. on model output rather than raw data. This is in itself not uncommon in fisheries science when computational or practical reasons prohibit integrated models that are based directly on raw data. This approach does, however, violate the assumptions of statistical analyses, so that care is needed. The splitting algorithm developed in paper I relies on covariates which affect the species ratio; for example, the depth, and the present study examines among other factors the variation with depth in the CPUE of each species. This double use of covariates, such as depth, implies a spill-over of uncertainty: an error in how the species ratio varies with depth will imply an error on the species splitting and therefore the species-specific catches, and this could then be interpreted wrongly as a depth pattern in the abundance of each species. It is difficult to quantify the uncertainty that arises from this double use of covariates, without conducting a major simulation study. It would be preferable to construct an integrated study that builds on raw data and which includes both species ratios and the CPUE. However, given the complexity and a large number of observations, this was not practically possible. Future improvement in computational abilities may make such an integrated model feasible.

3.3 Combined hake assessment vs single species hake assessments (Paper III)

Traditionally, and presently, the two species *M. capensis* and *M. paradoxus* off the Namibian coast are assessed combined as a single stock, using the Statistical catch-at-age analysis (Butterworth and Rademeyer 2005; Kirchner et al. 2012). The reason for this has been that fishermen do not distinguish the two species, but record the commercial catch data in the logbooks as one hake species. However, differences in biology and ecology suggest that the species should be managed and assessed separately. The work in Paper I and subsequent work in Paper II closes this gap between commercial catch data and species-specific assessments. Hence, in this paper, we perform the first species-specific assessments using the state-space framework. We apply three state-space models to assess hake stocks off Namibia: two independent single-species assessments, one for each species, and one combined assessment which treats the two species as a single stock.

This work was fully detailed and published in Environmental Development in 2016 as: *Kathena, J.N., Nielsen, A., Thygesen, U.H., Berg, C.W.*, 2016. *Hake species (Merluccius capensis and M. paradoxus) assessment in*
A brief description of the methods and highlights of the results are given below.

### 3.3.1 Materials and Methods

#### 3.3.1.1 Input data

In this paper, we use data for Namibian hake stocks for the period 1998-2012. We used the estimated numbers-at-age (ages 0-8 years) from surveys and commercial catches. Otoliths samples, from which the age information are derived, are collected annually during the summer swept area biomass survey. The maturity ogives for the two hake species are calculated annually for assessment purposes and to describe the species-specific age at maturity.

For the combined assessment, we need maturity ogives and weight-at-age, but we only have these for the separate species. The species combined maturities and weight-at-age were calculated as the weighted mean of the species-specific value, with weights being the numbers-at-age proportions by species as estimated by the species-specific assessments.

#### 3.3.1.2 Fundamental of SAM

Traditional single species fish stock assessment is based on cohort analysis, where the models track the number of surviving individuals in a cohort. This number decays according to the equation

\[ N_{t+1} = N_t \cdot e^{-(M+F)} \]

where \( N_t \) is the number of fish alive at time \( t \), \( F_t \) is the yearly fishing mortality and \( M \) is the natural mortality. Fishing and natural mortality are to be understood as average mortalities over the year, that apply to this cohort.
The number of individuals that die is equated to \( N_t - N_{t+1} \), and out of these a fraction \( \frac{F}{M+F} \) are fished while the remaining fraction \( \frac{M}{M+F} \) die of other causes.

Therefore, the catch of that cohort during the year is:

\[
C_t = (N_t - N_{t+1}) \times \frac{F_t}{(F_t + M)}
\]

A basic problem of fisheries assessment is that there are too many unknowns in these equations. The natural mortality \( M \) is really unknown but typically assumed known and constant. Even with this simplification, for each year and each cohort, the abundance \( N \) and the fishing mortality \( F \) are unknown, and only the catch \( C \) is known (assuming the data-rich situation where the age-structure of the commercial catch is known). This contrasts the desirable situation in statistics, which is that there are many more data points than parameters to be estimated. Survey data helps this situation by providing independent measurements of the abundances \( N \).

To reduce the number of unknowns even further, one approach is to assume a multiplicative structure, where the fishing mortality \( F_{a,t} \) on age group \( a \) in year \( t \) can be written

\[
F_{a,t} = F_a \times F_t
\]

The motivation for this structure is that fishing techniques remain constant, so that the relative mortalities on the different age groups do not vary between years, but that the overall fishing mortality \( F_t \) varies from year to year.

A more modern approach to assessment models is to make use of mixed effects models, where the abundances \( N \) and fishing mortalities \( F \) are not considered unknown parameters, but rather unobserved random variables. The model assigns prior distributions to abundances and fishing mortalities, and when the model is confronted with data, posterior distributions of \( N \) and \( F \) arise, from which the final estimates and confidence intervals are computed.
In SAM, this rule consists of the cohort equation, with some noise added in the logarithmic domain, while the fishing mortality for each year class is a random walk in the logarithmic domain. The motivation for the logarithmic domain is that uncertainty is relative; it also implies that the abundances and fishing mortalities necessarily stay positive. The random walks of the fishing mortalities may be correlated between age classes, with the same motivation as the multiplicative structure. The recruitment model needs to be specified, i.e. the abundance of the youngest year class. One may assume a stock-recruitment relationship, but SAM recognizes that the spawning stock biomass holds very little information about next year's recruitment, and therefore assumes that the each year's recruitment is simply a random variable independent of everything else.

3.3.2 Results highlights
The trends in the estimated spawning stock biomasses and fishing mortalities are similar between the species-combined assessment and *M. capensis* species-specific assessment (Figure 16 and 17). The spawning stock biomass for *M. paradoxus* followed different patterns than *M. capensis* (Figure 16). While the species-specific assessment for *M. capensis* shows an increase in the later years the abundance of *M. paradoxus* has remained more or less constant of the whole time-series. The Fbar for *M. paradoxus* is higher than the Fbar for the combined-species assessment (Figure 17).
Figure 16: Spawning stock biomass estimates with their associated 95% confidence interval as estimated by the three model assessment: the combined assessment and the two independent single species assessment. Included is also the two single species assessment added together (Figure 6 in paper III).

Figure 17: Estimated fishing mortality (Fbar: average over 3-6 ages) and their 95% confidence interval over time for the two independent single species assessment. For graphical clarity, the combined assessment estimate is not presented with confidence interval (Figure 7 from paper III).
3.3.3 Discussion
In this analysis, we compared the results of applying state-space species-specific assessments and a state-space species-combined assessment to Namibian hake (Merluccius capensis and M. paradoxus) data. The species-specific assessments showed that the spawning stock biomass of M. capensis has increased in the last five years while that of M. paradoxus has decreased. Fishing mortality on M. paradoxus is higher than fishing mortality on the combined-species. In the context of this study, the morphological similarities between the two species, and the combined fishery argue for a combined assessment, but differences in biological rates and spatial distributions speak for separate assessments of the two species. In this situation, it is important to understand quantitatively the differences in our perception of hake stocks that result from combined-species and split-species assessments. Ignoring the biological parameters of each species may lead to over-harvesting or under-harvesting of one stock.

Using the results from the species-specific assessments for creating input to the combined assessment, for example, maturity ogives, are clear examples of introducing circularities due to problems of estimating proper values of parameters for an artificial stock as the combined hake stock really is. The sensitivity of combined assessment results could be examined by robustness tests using alternative methods of parameter estimations such as average weighting based on overall catch proportions by species. However, it is difficult to avoid circularities completely, and hake splitting procedures, which clearly serve as the fundamental input to species-specific assessments, are still a challenge. Under these circumstances, it could be recommended to develop an annual reference case assessment for Namibian hake based on models that assess the two species as two independent stocks and is fitted to species-disaggregated data as well as species-combined data.

Similarly, for age-based stock assessments such as SAMs, correct age-length keys (ALKs) constitute a fundamental input to the assessments. Thus robustness testing of the impact of possible uncertainties in ageing procedures needs to be given priority, this is explored in the subsequent section.
3.4 The uncertainty of assessment due to ageing biases (systematic error)
In this section, we investigate the effect of age-reading bias on the assessment. Both age-reading bias and variance are a concern in age-structured stock assessment methods (Campana et al. 1995), but this study focuses on the bias. This problem has been identified in many fisheries worldwide (Campana et al. 1995). The ageing-reading bias affects all age-structured parameters in assessment models (Kalish et al. 1995). Age-reading is partly based on subjective judgments, where age-readers rely heavily on their experience. Several studies have reported problems of incorrect age readings mainly due to: lack of visibility of translucent rings for the Eastern Baltic cod (Reeves 2003); the otoliths nucleus region in older fish thickens, \textit{Lophius piscatorius} in the North Irish sea (Crozier 1989); and the opaque zones have a tendency of splitting in two in the case of \textit{L. vomerinus} off Namibia (Maartens et al. 1999).

Given the uncertainties in age reading and the central role of age data in age-structured assessment models, it is perhaps surprising that the literature contains only relatively few studies of the effect of age reading errors on assessments. Reeves (2003) simulated age reading errors on Baltic cod, including both variance and bias, and next simulated XSA assessment runs using the erroneous age data. The conclusion from that study was that the age reading errors did not affect stock significantly, but did affect catch forecasts and management advice. Bertignac and Pontual (2007) performed a similar study on hake (\textit{Merluccius merluccius}) in the Bay of Biscay and found broadly similar conclusions.

This section describes a simulation experiment which aims to examine the effect of biased age readings on the estimates produced by SAM on SSB’s, recruitment and fishing mortality.

3.4.1 Material and method
The study is based on a simulated data set which is simulated with SAM, where parameters are estimates based on Cape monkfish. This ensures that the simulation study is reasonably close to reality, but the study does not hinge on the accuracy of the assessment. We then perturb the data sets, to simulate the effect of bias in age reading. This results in perturbed numbers-at-age for both the commercial catches and the surveys shown in Figure 18. It also results in perturbed weight-at-age curves shown in Figure 19.
The data sets are perturbed as follows: We assume that all age-readers make a 10% error on the ages. We create two perturbed data sets in order to examine errors in both directions, i.e. first we examine the effect of under-estimating the age, then the effect of over-estimating the age.

For the case of under-estimation of the age, we assume that the read age is a random variable, which conditional on the true age follows a binomial distribution, where the true age is the number parameter and the probability parameter is 0.9. Thus, each true year ring has a probability of 0.9 of being detected by the age reader.

For the case of over-estimation of the age, the age-reader identifies features as year rings, even if they are not true year rings. We assume that the number of such false year rings is a random variable, which conditional on the true age follows a negative binomial distribution with probability parameter 0.9. The true and perturbed age distributions in the commercial catches are seen in Figure 18.

Next, we perturb the weight-at-age curves, assuming the same errors in the read age which form the basis of these curves. For example, if the age is over-estimated, the weight-at-age is under-estimated (Figure 19). We then run assessments, based on both the ”true” (unperturbed) data set and the perturbed data sets. We compare the results of these effects, to investigate the effect of age reading bias on the assessment.
Figure 18: True age data and the perturbed data: Mean commercial catches (top panel) and survey catches (bottom panel), of the simulated data sets. The “base” represents the true numbers-at-age. “Younger” represents the age distribution an age reader would construct if he under-estimates the age of a fish, while “older” represents the age distribution an age-reader would construct if he over-estimates the age of a fish.
Figure 19: Stock mean weight-at-age of the base run and the updated stock mean weight-at-age that are consistent with the assumed age-reading error. Base: represents the true weight-at-age. Younger: represents the growth curve an age reader would construct if he underestimates the age of a fish. Older: represents the growth curve an age-reader would construct if he over-estimates the age of a fish.

3.4.2 Results highlights

Figure 20 (a-c) shows the results of the base run and the runs with simulated age-reading errors. Over-estimating the age implies an over-estimation of SSB and under-estimation of fishing mortality. Under-estimating the age leads to the converse results, namely under-estimation of SSB and over-estimation of fishing mortality. Over-estimating the age leads to over-estimation of recruitment, while under-estimating the age leads to under-estimation of recruitment.

Here, we assume 10% error on the ages. When the ages are over-estimated by 10% the model produced a higher estimate of SSB roughly two times larger than the base run, while fishing mortality is twice lower than the base run throughout the whole time period (Figure 20 a-b). The estimates of SSB and fishing mortality are outside the 95% confidence interval of the base case. The recruitment trend is estimated to be very close to the base run and is within the 95% confidence interval (Figure 20c).
On the other hand, when the ages are under-estimated by 10% the model produced a low SSB and low recruitment than the base run, while fishing mortality is 1.5 times higher than the base run throughout the study period (Figure 20 a-c). The estimated SSB, fishing mortality and recruitment were all within the 95% confidence interval of the base run.

![Figure 20](image)

**3.4.3 Discussion**

We conducted a simulation experiment to investigate the effects of age-readings bias on SAM’s estimate of SSB’s recruitment and fishing mortality of Cape monkfish. The SSB estimates are up to almost two times higher than base run when the ages are over-estimated and lower than the base when the ages are under-estimated. Similar trends are observed in the recruitment estimations, but with different magnitude. The
estimated fishing mortality is two times lower than the base run when the ages are over-estimated and 1.5 times higher than the base run when the ages are under-estimated. These results indicate that errors in the age readings can have a significant effect on the assessment, even if temporal patterns in the time series are similar. These conclusions are qualitatively similar to those reached by Reeves (2003) and Bertignac and Pontual (2007). If the age is overestimated, immature fish are wrongly classified as mature, which will lead to an overestimation of the spawning stock biomass. This assumes that the correct maturity ogives are used. Conversely, underestimating the age will result in more fish classified as immature and hence reduce the spawning stock biomass.

The result highlights the impact that biases in age-reading can have on the estimates of management quantities. This also highlights the importance of ensuring that the bias in age-reading is reduced. On the other hand, these biases motivate analysts to continuously develop and improve assessment methods that do not rely on age-structured data, which is explored further in the subsequent section.

3.5 Data-rich vs data-moderate assessment of Namibian stocks (Paper IV)

“All models are wrong but some are useful”. – George Box

There is an increasing interest in assessment techniques that require less data than conventional methods. This interest comes from an aim of having more stocks assessed; also from stocks of lesser economic importance, or species that appear mostly as by-catch, for example, to be able to document that the fishery is sustainable. Data for conventional assessments require costly surveys, monitoring programmes, and laboratory analyses of samples to determine ageing and maturities, and therefore there is an interest in assessment modelling frameworks that require fewer data.

The aim of the present study was to investigate the applicability of a recent data-moderate assessment model, SPiCT, to Namibian stocks. A first part of the study compared the results from this data-moderate model with the results of a data-rich assessment method.

The Namibian government Marine Resources Act (Act No. 27 of 2000) Section 32 (2) obligates all fishermen harvesting marine resources in Namibian waters to record all harvest in the logbooks (MFMR2000), and
regularly submit these logbooks to the scientists at NatMIRC. The Marine Resources Act also stipulates that all by-catch species must be landed and recorded in the logbooks (MFMR 2000). The logbooks information is used as input to stock assessment methods, a tool for management advice. In this study, we utilize information from the commercial fishing operation targeting Cape monkfish. We supplemented this information with the scientific trawl survey also targeting Cape monkfish. We applied a biomass dynamic model to a data-rich Cape monkfish and a data-moderate west coast sole, the main bycatch in the Cape monkfish fishery. Surplus production models are simple, computationally-efficient methods to fit an observed time series of index of relative abundance (Martell and Froese 2013; Froese et al. 2017; Pedersen and Berg 2017). Surplus Production is simply the change in biomass ($B$) from one year to the next (Hilborn 2001). Surplus production ($SP$) arises through compensatory population processes that occur as a result of mortality, due to harvest (Prager 1994; Hilborn 2001). Thus, surplus production specifically incorporates biomass lost to fishing through yearly catch and are not differentiated by age and/or size (Prager 1994). In surplus production models, somatic growth, reproduction and natural mortality are not separately captured but consolidated in the two parameters: the intrinsic rate of population increase $r$ and carrying capacity $K$ (Froese et al. 2017; Polacheck et al. 1993). This is perceived as the greatest shortcoming of surplus production models. Recent development has grouped surplus production models as state-space models that separate process and observation errors (Buckland et al. 2004). Stochastic Surplus Production model in Continuous Time (SPiCT) (Pedersen and Berg 2017) is an example of one of such models. SPiCT is an updated version of the traditional Pella-Tomlinson surplus production model (Pella-Tomlinson 1969). The essential improvement of SPiCT over the tradition Pella and Tomlinson surplus production model is the implementation of a state-space approach (Pedersen and Berg 2017). Process error can account for natural variability of stock biomass due to stochasticity in recruitment, natural mortality, growth and maturation, whereas observation error determines the uncertainty in the observed abundance index due to measurement error and unaccounted variations in catchability (Carvalho et al. 2014; Francis et al. 2003).

This work was fully detailed and published in African Journal of Marine Science in 2018 as: Kathena, J.N., Kokkalis, A., Pedersen, W. M., Beyer, J.E., Thygesen, U. H. 2018. Data-moderate assessment of Cape monkfish
(Lophius vomerinus) and West Coast sole (Austroglossus microlepis) in Namibian waters. African Journal of Marine Science, 40(3): 293-302, https://doi.org/10.2989/1814232X.2018.1512527 This peer-reviewed paper is attached as paper IV in the present thesis (see Appendix IV).

A brief description of the methods and highlights of the results are given below.

3.5.1 Modelling and Data

3.5.1.1 Fundament beneath SPiCT

The most basic model for the dynamics of the total biomass is the continuous-time logistic growth equation (Prager 2002)

$$\frac{dB_t}{dt} = r B_t \left(1 - \frac{B_t}{K}\right)$$

where $r$ is the intrinsic rate of population growth, $B$ is stock biomass, while $K$ is the carrying capacity; if this stock is left alone unfished, the population will grow until it stagnates at $B=K$. If we add fishing mortality, we obtain the Schaefer model

$$\frac{dB_t}{dt} = r B_t \left(1 - \frac{B_t}{K}\right) - F_t B_t$$

whereas the Pella-Tomlinson generalization replaces the linear density dependence $\left(1 - \frac{B}{K}\right)$ with the more flexible $\left(1 - \frac{B}{K}\right)^n$ where $n$ is a shape parameter that determines at which $B/K$ ratio maximum surplus production is attained.

The SPiCT model (Stochastic Surplus Production model in Continuous Time (Pedersen and Berg 2017)) extends the Pella-Tomlinson model by including process error in the form of added stochastic noise $(\sigma_B B_t dW_t)$. Here $\sigma_B$ is the standard deviation of the process noise, and $W_t$ is standard Brownian motion. SPiCT estimates biomasses, fishing mortality and system parameters in a state-space framework, see Pedersen and Berg (2017) for a fuller description of the model.
SPiCT is formulated in continuous time allowing the model to handle varying sample times. For example, the two input indices, commercial landings and survey indices may be recorded at any time of the year and still be used as data for the estimation process. Input data to SPiCT are total catch and catch per unit of effort from fishery dependent and fishery independent sources. SPiCT assumes that fishing mortality acts on the total exploitable biomass and neglects population structure and its impact on fisheries selectivity and maturation.

### 3.5.1.2 Comparison with age-structure data

The comparison of biomasses between the two methods was done by calculating a normalized error. The formula used to calculate a normalized error is wrongly published in Paper IV and should be published as:

\[
\frac{\log(B/B_{MSY})^{SPiCT} - \log(B/B_{MSY})^{SAM}}{\sqrt{\sigma_{SPiCT}^2 + \sigma_{SAM}^2}}
\]

Where \(\sigma_{SPiCT}^2\) and \(\sigma_{SAM}^2\) are the variance of the estimates (in log domain) as reported by the two models. A similar error was computed for the fishing mortalities.

### 3.5.2 Result highlights

The SPiCT assessment for Cape monkfish produced a higher estimate of \(F/F_{MSY}\) than SAM in the middle of the time period (Figure 21a). The 95% confidence intervals for SPiCT were relatively large and exceeded those of SAM for most of the time period, except around 2003, where the confidence intervals do not overlap. After 2011, the two models presented different biomass trends: SAM showed declining biomass with increasing fishing mortality, whereas SPiCT showed stable relative biomass and a stable relative fishing mortality (Figure 21a, b). Measured by the normalised error, SPiCT and SAM gave largely consistent estimates of \(F/F_{MSY}\) except for a large difference peaking around 2003–2004 (Figure 22a). The \(B/B_{MSY}\) estimates match closely from 2003 to 2011, but SPiCT estimates higher relative biomass beyond these years, with the normalised errors exceeding 2 in the first and last years (Figure 22b).
Figure 21: Assessment of Cape monkfish \textit{Lophius vomerinus} in terms of (a) relative fishing mortality $F/F_{\text{msy}}$ and (b) relative to exploitable biomass $B/B_{\text{msy}}$; the y-axes are in log scale. Two methods were used: state-space assessment model (SAM) (solid lines, the dark grey area represents 95% confidence bounds) and the stochastic surplus production model in continuous time (SPiCT) (dashed lines, the pale grey area represents 95% confidence bounds (Figure 2 in paper IV)).
3.5.3 Discussion

In this study, we evaluated the data-moderate assessment method SPiCT, applied to Cape monkfish, by comparing its results with a SAM assessment of the species. We provide for the first time a SAM assessment of Cape monkfish in Namibian waters. Furthermore, using SPiCT, we provide the first assessment of the west coast sole stock off Namibia, which is currently unassessed. The results of SPiCT and SAM yield largely consistent assessments of Cape monkfish, even if there are noticeable discrepancies.

In these situations, a relevant question is which of the two models is more credible. While the biomass dynamics of SPiCT is a fairly coarse description of population dynamics, in that it ignores sex-, size- and age-based differences by treating a stock as undifferentiated biomass, the higher fidelity of SAM hinges on the accuracy of the age readings, which is doubtful. We are therefore not yet able to reliably assign relative
credibility to the two models. To remedy this situation, future studies could explore additional validation techniques for these models, as well as analyse the sensitivity of the models, such as through simulation studies.

Age information is both costly to acquire and prone to errors, which explains the current interest in alternative assessment methods that do not require such age information, namely data-moderate methods. This is particularly relevant for Namibian stocks that are currently unassessed. Although the present study has identified some differences between estimates from SAM and from SPiCT, the potential of data-moderate methods justifies further studies of their performance, and their particular use for assessing the Cape monkfish and west coast sole stocks in Namibian waters—thereby supporting informed management decisions. We consider this a promising start for applying data-moderate assessments to Namibian fish stocks.

It should be noted that length data for monkfish and sole exist from commercial fishing, collected onboard by the observers from 1998 (1996 for monkfish) and from the scientific surveys from 2000 onwards. This opens up for applying promising new length-based stock assessments (Rudd and Thorson 2018) not only for monkfish and sole but also in other considerably more data-limited fisheries where neither reliable ageing data, fishery-independent survey or catch time-series data is available. With a general understanding of biological parameters using the concept of Beverton-Holt life-history invariants (Prince et al. 2015), it is possible to give an idea of the status of a stock from a length distribution of fish in the catch from just one year assuming equilibrium conditions (Kokkalis et al. 2015; Hordyk et al. 2015; Rudd and Thorson 2018).
Chapter 4: Stock assessment perspective for Namibia

Stock assessment perspectives are introduced by considering hake catch splitting, single-species hake stock assessments and transboundary challenges in section 4.1 and data-limited fish stock assessment in section 4.2. On this basis I enter the conclusion in section 4.3 starting with summarizing the contributions of the thesis according to the objectives and after re-visiting hake challenges in Namibia from a biological perspective, I finally move towards improved assessment, advice, and management in the Benguela Large Marine Ecosystem (BCLME, section 4.3.4). I end up with a general statement that advances in fishery management and advice require spatial understanding of fisheries, fish stocks and their interactive dynamics. I see such a development towards increased spatio-temporal mechanistic knowledge of the vital rates as forming a fourth trend in fishery stock assessments.

4.1 Stock assessment perspective for Cape hake

4.1.1 Hake splitting

The only source of hake species separated data from commercial fishing is length frequency data collected by fisheries observers’ onboard commercial fishing vessels. Any error in species identification by the observer during sampling would affect the species ratio, and that will propagate through the splitting algorithms into the assessment.

Splitting commercial catches of mixed hake into species is a crucial step in improving hake stock assessment off Namibia. This work provides a step forward advancing the Namibian hake assessment from a traditional combined assessment to a species disaggregated assessment. Species disaggregated numbers-at-age in the commercial catches are key inputs into the stock assessment model. This study did not find an indication to suggest that any bias is introduced by the method of splitting commercial catches.

The study recommends an intensive training programme for fisheries observers to ensure that species identification is fully understood. Furthermore, in the future, an integrated modelling should be considered to avoid a two steps approach currently applied in the analysis performed in Paper I and Paper II. The catch
species ratio should be based on analysis of a dataset preprocessed with an observer data model using also year and month as an explanatory variable. It is also worth noting that the species-disaggregated assessment should be updated into historical and most recent years. Future studies need to investigate the appropriate age-specific catchability and their implication to management quantities of interest. The study also recommends that CPUE standardization should consider including vessel ID, hake size classes and possible interaction effects using GAM or GLMMs. Including robustness tests, such comprehensive analyses require and deserves a PhD study by itself. Thus, species splitting of hake still constitutes an extremely important task to attend to in the future to continue improving hake specific stock assessments in Namibia.

4.1.2 Stock assessment
Data-rich stock assessment models heavily rely on age-structured data. Age-structured data are generated through reading annual rings on otoliths (Wilhelm et al. 2015). The survey and commercial catch-at-age information derived from the observed otoliths rings are used to estimate the stock size and fishing mortality (Wilhelm et al. 2015). Misreading of the annual rings would either lead to an underestimation or overestimation of the true age of that specific species.

The species disaggregated assessments showed that *M. capensis* has a higher spawning biomass than *M. paradoxus*, while the fishing mortalities are higher on *M. paradoxus*. This indicates that fishing pressure is higher on *M. paradoxus* and this can lead to local depletion if it is not addressed. This result appears to add a new dimension to the understanding of the hake species-specific stock status in the Northern Benguela. *M. paradoxus* has a late onset maturity and a lower intrinsic rate of natural increase, as compared to *M. capensis* (Paulus et al. 2013). In such a case a sound management strategy would be to take the smaller of the two $F_{MSY}$ estimates (Sparre and Venema 1998).

The results of this study have increased knowledge and understanding on age-specific catchability i.e. systematic patterns in the residual plots of the younger age classes were observed, but the situation improved when the younger age classes were allocated a different standard deviation indicating that assuming constant catchability by age is inappropriate.
In addition, based on recent descriptions of hake spawning peaks from growth rates and survey information (Wilhelm et al. 2013, 2015), from the gonadosomatic index (Jansen et al. 2015) and given length-at-maturity from histological analysis (Singh et al. 2011), adjustments need to be made in the current assessment of the timing of recruitment and the length at maturity for both *M. capensis* and *M. paradoxus*.

Results from the age-reading error experiment in this study indicate that errors in the age reading can have significant effects on the assessments. This highlights the importance of controlling age-reading. However, it also highlights the importance of robustness tests concerning assessment consequences of uncertainties in splitting and ageing. Length-based input as an alternative to a purely age-based input to the assessment may also be considered in the future together with realistic stochastic growth modelling.

4.1.3 Transboundary challenges

Several publications have shown that the spatial distribution of both hake species extends into South African waters. More recent evidence emanating from the Ecofish project has reported that *M. capensis* in Namibian waters is a unit stock whereas *M. paradoxus* forms a transboundary stock extending from Namibian waters towards the southern Benguela and is hence shared with South Africa (Strømme et al. 2016; Jansen et al. 2017; Henriques et al. 2016). The situation underscores the need for transboundary management of the Cape hake stocks. This evidence brings to the fore a complication of two completely different problems facing the current stock assessment of hake in the Namibian water i.e. assessing the two species combined as a single stock and then assuming that the *M. paradoxus* in Namibia is a distinct stock independent from the South African part of the population. The evidence provided suggests that current assessments in Namibia and South Africa are not assessing the *M. paradoxus* stock in its totality.
Figure 23: Time series of Namibian share of *M. paradoxus* biomass based on the surveys in the years 2005–2012, and arithmetic mean for all these years. The lowest and highest estimates are in 2010 and 2012, respectively. Adult range 35-55 cm in Namibia indicated. Adopted from Strømme et al. (2016).

The time series of the Namibian share of *M. paradoxus* presented by Strømme et al. (2016) shows this very clearly, see Figure 23. A remarkable stable plateau shows that on average 40% of the adult *M. paradoxus* stock in the size range 35 cm to 55 cm is found in Namibia. These fish are recruited from South Africa at around 20 cm since no evidence of spawning has been detected in Namibia. *M. paradoxus* do not grow bigger in Namibia either because they are fished or because they return to South Africa. As mentioned above, the evidence of homing behavior is piling up. This catch picture becomes even sharper when the effect of size correlations etc. are incorporated in geostatistical modelling and the impact of local noise (nugget effect) removed (Figure 10c in Jansen et al. 2017): virtually all 60+ cm fish stay south of the border. It is difficult not to conclude that large fish eventually return to the spawning grounds in South Africa, but the frequency and strength of this migration is still not known (Strømme et al. 2016).

As an example of elucidating this migration effect on the Namibian stock assessment of *M. paradoxus* we may consider natal homing as a continuous process described by an instantaneous rate of migration (*L*γ,a), (L for leave) such that cohorts annually are not only reduced by the standard exponential survivorship, exp(-(*M*y,a +
F_{y,a}) as described in section 3.3.1.2, but also by exp(-L_{y,a}). Thus the impact can be examined by replacing natural mortality, M_{y,a} with M_{y,a} + L_{y,a}. One problem, of course, is that we do not know L_{y,a} but neither do we actually know M_{y,a}. What we expect is that M due to predation decreases with size or age and that L increases with age. If L is ignored as in Paper III one could expect an artificial higher F estimate for adult fish, but further speculations here would benefit from simulation experiments as suggested in closing Paper III.

### 4.2 Data-limited fish stock assessment

SPiCT has provided a chance to assess data-moderate stock and make inferences. It has therefore provided an opportunity to investigate and contribute to the understanding of species-specific dynamics of by-catch species. Estimated species-specific fishing mortality and spawning stock biomass in both data rich and data poor stocks make it easier to provide species-specific management advice. Understanding the temporal development in fishing mortality is one of the important steps in fisheries management.

Although the study reveals the status of *A. microlepis* stock, more data are needed in order to help establish effective management advice for this stock. This is particularly important because *A. microlepis* is an important by-catch species in monk fisheries and exploitation of this stock will continue for many more years to come. The results from this study will also be used to explore the assessment of other by-catch demersal living resources off Namibia that are currently unassessed. The SAM and SPiCT toolboxes are continuously under development and a new SPiCT-version will be ready in 2019, this version is able to account for changes in productivity over time both within and between years (Casper W Berg, personal communication). It has already been used to demonstrate a regime shift for Baltic cod which raises questions about the concept of a single MSY. It will be interesting to use this model on hake, for example, based on long time series of combined hake catches.

It is not only the amount of data that should determine to which degree a fishery is data-limited. The quality of the data should play an equally important role. For this and several other reasons, the development of data-limited methods for assessing the status of a stock seems to continue. Data-limited methods could be part of the assessment for all stocks. This study has focused on SPiCT, but in particular, length-based methods should
receive attention. Length-based assessment methods constitute an important part of data-limited methods because it is much easier to obtain reliable length measurements of catch samples than to measure total catch or record effort data for small-scale or non-target species (Harley et al. 2001; Kokkalis et al. 2015; Rudd and Thorson 2018).

4.3 Conclusions
This final section of my thesis moves from summarizing the results according to the specific objectives given in section 1.2 to some questions of how we in the region can continue improving this new line of implementing and interpreting state-space assessment techniques such as SAM and SPiCT. Then in section 4.3.3, realizing that the machine room of SAM and SPiCT apart from its advanced statistics is still based on the classical assumptions of fishery biology outlined in paragraphs 3.3.1.2 and 3.5.1.1, I return to biological processes and the need for improving current knowledge of the vital rates. Finally, moving from short-term to longer-term goals, I consider my ultimate goal of facilitating improved assessment, advice, and management in the Benguela Current Large Marine Ecosystem.

4.3.1 Summarizing according to the objectives
The thesis contributes with research providing new and more precise population dynamic parameters improving the knowledge basis for the fisheries biology of demersal stocks off Namibia. It provides new input to the foundation and input to hake stock assessment. This includes developing the first existing method for splitting the hake species in the combined hake landings, more advanced approaches including evaluating the impact on assessments due to systematic ageing bias, as well as new methods with GAMs to provide standardized catch rates with species-specific abundances at each commercial haul and to identifying factors that are affecting catch rates or CPUE for each tow. On this basis, the thesis provides an application of the new parameters and information in stochastic assessment methods to test the assessment performance and applicability according to the new input data. The results are evaluated and discussed by comparing single stock runs with combined hake runs with the same stochastic assessment model (SAM). A sensible next step could be to compare assessment models for the hake stocks and their advisory and management implications.
but that was not part of the objectives, which is why a scientific report from 2015 comparing SCAA and SAM modelling although interesting is not discussed in the thesis.

Finally, to assess a data-poor Namibian stock for the first time the new stochastic version of surplus production modelling (SPiCT) was applied to west coast sole. SPiCT appears to be a sensible choice when ageing data are not available or trustworthy because without using available ageing information it has proved its value in a number of ICES cases when compared to the results of age-based SAM assessments. For monkfish both SAM and SPiCT were applied to evaluate the performance of the latter and its application to data-poor stocks in general in the Benguela Current Large Marine Ecosystem (BCLME).

Although the study reveals the status of west coast sole stock, further investigations are necessary in order to help establish effective management advice for this stock. One problem is that current time-series are too short for reliable estimation, and naturally this problem will gradually be resolved with continued data collection. The results from this study will also be used to explore the assessment of other by-catch demersal living resources off Namibia that are currently unassessed. We consider this a promising start for applying data-moderate assessments to Namibian fish stocks.

To improve these very first individual assessments of *M. paradoxus* and *M. capensis* in Paper III my first tasks will be to build on the knowledge gained from my PhD study including the ageing-bias examination in order to quantify the uncertainties of the species-specific catch and survey time-series input. This includes for example new attempts to prolong the time-series hopefully back to the start of surveys in 1990 with attached uncertainties, re-analysis of Namibian hake survey data removing time of day effect, and identifying anomalous years potentially due to environmental forcing such as 2010 where the Namibian ratio of adult *M. paradoxus* biomass suddenly was halved (Figure 23). These analyses will also involve updating knowledge on vital rates (section 4.3.3) and other preparatory tasks required for the implementation of transboundary *M. paradoxus* stock assessment (section 4.3.4). The successful improvement of stock assessments further relies on Namibian fishery biologists being able to take advantage of new developments of the toolbox as indicated in the next section.
4.3.2 How do regional fishery biologists learn to operate state-space assessment techniques?

My PhD study was initialized as part of the Ecofish action since it was realized that it would be hard work to create local ownership of modern stock assessment techniques. As a biologist, I can only agree. I realize now that I must struggle the rest of my working life to continue communication with experts of the machine room and disseminate it to local colleagues. Such communication is necessary as tools and possibilities are developing all the time.

Regionally, there are BCC Working Groups established as a result of the Ecofish project aimed to set-up a joint assessment. The general aim of the Working Groups is to contribute to the improved management of shared demersal and pelagic resources in the BCLME through cooperative work among member countries in order to ensure sustainable use of these resources. The Regional Working Groups Terms of Reference make provision to establish Task Teams to address specific technical issues as required. Under this specific function, the working group would invite external experts and that is one way the region can continue improving this new line of implementing and interpreting state-space assessment techniques such as SAM and SPiCT. Parallel, regional fishery biologists can learn through a variety of mechanisms such as personnel exchanging of stock assessment scientists, or cooperative research projects, in-service training courses offered on-site, or professional leave to take university courses, specialized short courses, and workshops.

These combined activities should ensure the necessary continuity in learning and training and I hope personally to be able to contribute to such cascading effects for the region.

4.3.3 Challenges from a fishery biological perspective: Cape hake

It is important for stock assessment and management to get the vital rates right but more difficult than I anticipated at the start of this study. Concerning recruitment, for example, the case of Namibian *M. paradoxus* adds the complexity of spawning not taking place off Namibia. In this respect, Namibian *M. paradoxus* is to be considered an adult extension of the South African stock (Strømme et al. 2016) so recruitment in Namibia is most likely entirely determined by what happens in South Africa.
Reading age rings on hake otoliths is still a challenge and if length-based modelling is considered an alternative we do not really know how individual growth takes place and how much it differs annually in space and time. Just assuming a one-to-one correspondence between age and some average VBGE-growth curve is not solving the problem, but transferring unknown ageing to an unknown growth problem. Since available knowledge on growth, in general, indicates that individual growth trajectories do not overlap but tend to spread out it is a well-known procedure to describe individual growth with the same curvature parameter K of the VBGE but with individual Linf following a log-normal distribution or similar (Beyer and Lassen 1994).

Without correct ageing, it is difficult to get mortalities right in the classical cohort approach underlying SAM. In the case of Namibian M. paradoxus, we are probably now also facing migration at least of bigger fish back to South Africa. Simulation experiments with migration as an additional natural mortality component as suggested in section 4.1.3 could gain insight into the nature of migration that would have a profound effect on the Namibian assessment of the species.

Understanding and quantifying natural mortality constitutes a major challenge and it will, in particular, be important to gain insight into hake-on-hake predation and cannibalism. Hake stomach sampling has been initiated in Namibia but needs further pursuing (Kainge 2017).

4.3.4 Towards improved assessment, advice, and management in the Benguela Current Large Marine Ecosystem (BCLME)

The assessment models considered in this thesis provide the short-term predictions that are required for yearly management decisions such as setting total allowable catches. More strategic management decisions require trustworthy long-term predictions, which include the response to changing climate, ecosystem composition, and species interactions. For such purposes, proper representation of the mechanisms of stock dynamics (reproductive success, growth, mortality including both cannibalism and multispecies interactions, and migration) becomes more critical. As such, although the work in this thesis has contributed to the scientific basis for management of important demersal living resources off Namibia and in BCLME, it should be clear from the previous discussion and my four Papers that the work should be continued in several directions. Focus in ending this discussion will be on transboundary challenges.
BCC was established to monitor the major transboundary fish resources in the region. A primary candidate resource for such consideration is hake (*Merluccius capensis* and *M. paradoxus*) off the coast of Namibia and South Africa and horse mackerel (*Trachurus trachurus capensis*) off the coast of Namibia and Angola. A total of 9 trans-boundary surveys were conducted between 2003-2012, monitored at the beginning of each year (January–February), to allow for a synoptic overview of distribution and abundance of resources that may be shared across the region. The transboundary hake challenge has been a subject of discussion for a long time.

Shared-stocks and management issues for marine fish occur virtually throughout the world, and there tends to be a natural progression in how potentially shared-stock fisheries proceed. Typically, cooperation begins in the area of scientific research through the development of common sampling protocols.

Shared-stocks research leading to effective management is a priority for Namibia at the moment, since the Namibian hake fishery is being assessed by the Marine Stewardship Council (MSC) for certification. MSC has requested Namibia to put in place a mechanism for organized and effective cooperation with South Africa, required by shared stocks, which can deliver outcomes for any shared hake stock. The South African hake fishery has already been certified and will be subjected to similar conditions in the next round of certification. Ideally, a number of different assessment modelling approaches should be attempted to make an assessment for the entire distributional area of *M. paradoxus*. It is worthwhile to mention that spatial box modelling was believed to constitute a promising and important approach in Ecofish, but it turned out after considerable trials for South African waters that too many parameters necessarily became involved, so it was not feasible to get useful results and South Africa stayed with their old fleet-as-area approach. The most promising tool developed was the geostatistical model GeoPop but the problem here is that the results of this model are only based on correlations from survey data.

Other developments in mixed-effect modelling create new possibilities for advanced stock assessment. On the state-space front, Namibia could benefit from a step-by-step development using SAM and SPICT. It is also very important that Namibia work towards improving the hake-splitting procedure in-cooperation with South Africa. The countries clearly need to agree on a number of procedures at various levels e.g. harmonization of data collection protocols.
SAM and SPiCT comparisons with SCAA will be needed. For further comparative details of the two methods applied to the Namibian hake stock in the case of combined species, the reader is referred to the report listed under other work performed during the PhD study (Kathena et al., 2015). It should be noted that SAM-SCAA comparisons have not yet been reported for species which do not have the complexities associated with those for Cape Hake, e.g. splitting the species challenge. However, such an international study should be in progress following the World Fisheries Congress in Korea in 2016. Results from such comparisons can be used to direct survey programs to improve data quality and to assess the degree of improvements achieved over time.

Building on the innovative hake splitting procedure the most important contribution of this PhD project is a promising start for applying stochastic assessment methods to data-rich as well as data-moderate fish stocks in the region. The published results serve as starting points for improving fish stock assessment and management in Namibia and elsewhere. Apart from assessment and management challenges of shared-stocks important challenges ahead include understanding the effects of species interaction, climate change and density-dependent mechanisms. This understanding could begin with elucidating the impact of species interaction on stock assessments in particular for Cape hake since *M. capensis* and *M. paradoxus* interact through hake-on-hake predation, food competition etc. Focus could be on connecting single-stock assessments through correlated survival (Albertsen et al. 2017). The strategy of analysis could be to first consider only hake interaction, then to add seals and later include other species (e.g monkfish) in the modelling. We may also learn something on the historical development from a combined hake assessment using the entire catch series from 1963 (Figure 5) since SPiCT now can deal with estimating regime shifts which are becoming of increasing importance under climate change (Mildenberger et al. 2019). What we lack in general to develop fishery management and advice a major step further most likely is the spatial understanding of fisheries, fish stocks and their interactive dynamics. I believe this understanding will increase tremendously once stock assessment tools are developed further based on a spatio-temporal mechanistic knowledge of the vital rates. We need to be able to quantify and scale the effects of local density-dependent mechanisms to understand important phenomena such as when in life recruitment is determined for various species. I see this development in an optimistic view because of promising starting points, tools such as GeoPop, together with fast on-going
progress in computer science and analysis (including TMB developments) most likely is about to form a fourth trend in fishery stock assessments.
References


Appendix I

Paper I

A robust method for generating separate catch time-series for each of the hake species caught in the Namibian trawl fishery.

Full citation:

Due to copyright, the paper is not included in this version of the thesis.
Appendix II

Paper II

Population abundance and seasonal migration patterns indicated by commercial catch-per-unit-effort of hakes (*Merluccius capensis* and *M. paradoxus*) in the northern Benguela Current Large Marine Ecosystem.

**Full citation:**


Due to copyright, the paper is not included in this version of the thesis.
Appendix III

Paper III

Hake species (*Merluccius capensis* and *M. paradoxus*) assessment in the Benguela Current Large MarineEcosystem

Full citation:


Due to copyright, the paper is not included in this version of the thesis.
Appendix IV

Paper IV

Data-moderate assessment of Cape monkfish (*Lophius vomerinus*) and West Coast sole (*Austroglossus microlepis*) in Namibian waters.

Full citation:

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