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Drop shape analysis for determination of dynamic contact angles by double sided elliptical fitting method

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Abstract

Contact angle measurements is a fast and simple way to measure surface properties and is therefore widely used to measure surface energy and quantify wetting of a solid surface by a liquid substance. In common praxis contact angle measurements are done with sessile drops on a horizontal surface fitted to a drop profile derived from the Young-Laplace equation. When measuring the wetting behaviour by tilting experiments this is not possible since it involves moving drops that are not in equilibrium. Here we present a fitting technique capable of determining the contact angle of asymmetric drops with very high accuracy even with blurry or noisy images. This we do by splitting the trace of a drop into a left and right part at the apex and then fit each side to an ellipse.

Keywords

Contact angle, wetting behaviour, roll-off angle, drop shape analysis, axisymmetric drop shape analysis (ADSA), triple-line, and contact angle hysteresis.
Measurements of contact angles between liquids and solids are widely used to determine properties of either the liquid or the solid. [1] When measuring the contact angle it is only the chemical properties of a few of the outermost atomic layers in the solid that affects the liquid, this makes contact angles a very simple way to measure surface properties. [2] The simplest and most used method to measure the contact angles of a drop is by depositing a drop of liquid on a solid surface and acquire a digital image of the drop in profile. [3] The image is then analysed to extract the coordinates for the drop profile and determine the position of the solid-liquid interface. To extract the contact angle from the obtained data the drop profile is fitted to an equation, which is evaluated at the triple-line. For a drop sitting on a horizontal and homogeneous surface, we can assume that the drop is axisymmetric around the vertical axis and the drop shape is therefore completely described by the hydrostatic Young Laplace equation. Fitting to the Young Laplace equation is called axisymmetric drop shape analysis (ADSA) and was at first limited to drops where the apex of the drop was visible[4] and have later been improved to be able to fit drops with only parts of the drop shape being visible (ADSA-No Apex). [5] It is widely accepted within the field of measuring contact angles that fitting to the Young Laplace equation provides measurements with the highest possible accuracy and improvements to the technique only concerns the fitting algorithm, the determination of the baseline and the exact position of the drop perimeter. An example of a different approach to fit the acquired image to the Young Laplace equation is Theoretical Image Fitting Analysis (TIFA) [4, 6] where theoretical images are generated and compared to the acquired image, thereby circumventing the need for an edge detection algorithm. The validity of the Young Laplace equation is however limited to symmetric drop shapes. This means that for measurement of the dynamic contact angles by tilting experiments there is a need for a different equation to fit the drop perimeter to. Several examples of this are derived approximations to the Young Laplace equation assuming some out of plane shape of the drop[7] or purely arbitrary equations like, cubic splines[8] or polynomials.[9, 10]

When choosing fitting algorithm there are two important properties that should be considered; firstly, the amount of data-points that can be fitted to the equation. This usually involves both a minimum number of points to achieve the desired precision and a maximum number of points where the equation is a good approximation to the drop shape. Secondly, the ability to extrapolate the drop shape outside the region of fitted data points, since optical distortions at the triple-point require the fitted equation to be extrapolated down to the baseline.
For the axisymmetric case, the Young-Laplace based fitting methods are able to use the whole perimeter of the drop while being very accurate at extrapolating the drop shape since it is derived from the physical properties of the drop. For tilted drops, the most commonly used method is polynomial fitting due to its simplicity. According to Weierstrass approximation theorem polynomials can be as good a fit to a continuous function on a closed interval as desired, [11] this means that polynomials always will be able to fit parts of the drop shape, even for oddly shaped drops impacting a surface or under influence of electric fields etc. Polynomials are usually used as interpolants, where higher order or piecewise polynomials in general can be used to get results that are more accurate. For fitting drop shapes and measuring contact angles, the polynomial will however be used to extrapolate the drop shape making the degree of polynomial a trade-off between the maximum amount of points that can be used for the fit and the accuracy on the extrapolated drop shape. This makes polynomial fitting very sensitive to noise in the image and especially to blurry edges or optical defects at the contact point. In order to be able to fit to all data points on the drop perimeter we have found that fitting advancing and receding sides of the drop separately with two ellipses gives very accurate results for most real drops. This can be seen as a generalization of the concept presented by El Sherbini et.al. [7] where they fit a vertically inclined drop to two circular segments divided at the apex of the drop.

To be able to compare our elliptical fitting method to the more common polynomial fitting we have implemented a polynomial fitting algorithm together with the elliptical fitting method. In this way, we can ensure that the extracted drop perimeter and baseline detection is the same and that the difference in results only comes from the difference in fitting method. In our implementation of polynomial fitting we rotate and translate the data points for each side of the drop so that the curve of data points \((x, y)\) fulfil \(\frac{dy}{dx}_{x=0} = 0\) and \(\langle x \rangle = 0\). In this way we have good conditions for the polynomial fit regardless of the contact angle of the drop and thereby circumvent the difficulty in fitting polynomials to nearly vertical profiles for drops with a contact angle close to 90°. For implementation of polynomial fitting, it is required to select the degree of the polynomial and the amount of points used in the fit. We choose this by simulating points on circular segments with a small scatter and fitting the points to polynomials of various degrees. We found that fourth order polynomials are good for extrapolation of the slope while still being able to fit a large arc of the segment with high accuracy. These simulations are explained in detail in supplementary materials.

The drop shape analysis has been implemented in MATLAB (R2016a) and can be broken down into a series of steps, each shown in Figure 1. Step a), extract the perimeter of the drop. This is done using the algorithm
and script presented by Trujillo-Pino et al. [12] The method of Trujillo-Pino et al. provides edge detection with subpixel accuracy that is very similar to those obtained by sigmoidal fitting [10] of the edges while being computationally faster and more accurate at points where edges are close to each other, e.g. at the triple-line for very high or very low contact angles. Step b), determine the baseline by finding the reflection and then calculating the intersection between linear fits made to the data points just above and below the reflection. This should preferably be done on several recorded frames and then averaged to obtain a precise positioning of the baseline. Step c), if using polynomial fit, we need to select the amount of points needed for the fit. For this, we use a geometric relation for circular drops, that the arc between apex and contact point equals the contact angle. To select $s$ out of $n$ data points on the perimeter of the drop corresponding to an arc $\alpha$ with a drop contact angle $CA$, we use the relation $s = n \cdot \frac{\alpha}{CA}$. If the selected arc is smaller than the contact angle, all data points from apex to triple-line are used. For all polynomial fits presented in this paper we use data points in an arc of $\alpha = 60^\circ$, see supplementary material for details. Step d) fit the obtained data to the function of choice. We show both polynomial and elliptic fits. Polynomials are fitted using standard linear least squares fitting whereas ellipses are fitted using the direct elliptic fitting method proposed by Fitzgibbon et al. [13] Step e), evaluate the slope of the fitted function at the intersection between fit and baseline. From this, the contact angle is calculated.

Figure 1 Step-by-step implementation of the contact angle fitting algorithm. (a) Edge detection. (b) Baseline detection. (c) Selection of data points used in polynomial fit. (d) Elliptic and polynomial fit. (e) Evaluation of fit at the intersection with baseline to obtain the contact angle.
To evaluate the implementation of both the double-sided elliptic and the polynomial fitting algorithm we have constructed a series of synthetic images. We constructed all synthetic images to resemble images obtained by our goniometer (Attension theta, Biolin Scientific) but with well-defined contact angles. After validating the numeric implementation of the fitting algorithms, we use the generation of synthetic images to determine the sensitivity of the algorithms to different kinds of distortions. Each synthetic image is generated by plotting circular segments with a smooth transition from black (drop) to white (open space). This is implemented by calculating the position of centre of the circular segment corresponding to a specific contact angle, in Figure 2 we have sketched a synthetic drop where the centre of the circular segment is positioned in $(0, D)$ where $D = -R \cos CA$, the radius $R$ is chosen so the drop can fit into the image. The drop edge is generated by evaluating a cumulative distribution function of a Gaussian distribution $F(s) = \frac{1}{2} \left[ 1 + \text{erf} \left( \frac{s-R}{\sigma \sqrt{2}} \right) \right]$ where $s$ is the distance to $C$, $R$ is the radius of the drop and $\sigma$ is the standard deviation describing the width of the greyscale transition. By analysing the goniometer images presented in this paper we find that the average width of the transition is $\sigma = 0.85 \pm 0.19$ pixel, and unless otherwise stated we have used $\sigma = 0.85$ pixel to generate the synthetic images.

Figure 2 Schematic drawing of the geometry used to generate synthetic images.

By generating synthetic images with contact angles ranging from $10^\circ$ to $170^\circ$ and fitting the images with both polynomial- fitting and elliptical fitting we obtain the data presented in Figure 3. From this, we see that the rotation of the drop boundary for polynomial fits ensures high accuracy also for contact angles near $90^\circ$. When comparing the performance of our implementation of polynomial fit with the results presented by Chini et al. [10] we see slightly worse performance for our program for the synthetic image resolution used; this can be improved by generating synthetic images with higher resolution (see the
Supplementary Material). Instead, we choose to evaluate the accuracy of our fitting technique by generating as realistic drop images as possible in bitmaps with a resolution of 512x337 pixels and a black to white transition width of 0.85 pixels. For almost all contact angles, the elliptic fit determines the contact angle more accurate than the polynomial fit, this is to be expected since the synthetic drops are generated as circular segments that are fitted perfectly by ellipses. The seemingly stochastic variation of the error on the determined contact angle is directly related to the chosen resolution of the synthetic image. When generating the synthetic images there will be a loss of information due to digitizing of the geometric shape into pixels. The error due to this information loss is directly linked to the exact digitizing of the drop and will therefore be different for different resolutions (see the Supplementary Material).

In the generation of synthetic images, it is directly possible to vary the sharpness of drop edges and add noise to the generated image. By changing the width of the greyscale transition in the image generation, we evaluate the error introduced by blurry images, for instance if the drop profile is taken slightly out of focus. We have varied the width of the greyscale transition from 0 pixels (completely sharp) to 3 pixels (blurry to the naked eye) of a drop with a contact angle of 140° and presented the error in contact angle in Figure 3b. For most synthetic drop shapes in the parameter space investigated, except for very low contact angles, where the elliptic fits are very sensitive to choice of the drop centre, and very noisy images, we see that the elliptic contact angle measurement is more accurate than the polynomial fit. It is, however, also apparent that the two graphs follow the same trend (on a log scale) indicating that the main error arises from the profile extraction that is the same for both drops. The minimum error for both fitting methods around $\sigma = 0.5$ is a consequence of the edge detection algorithm that utilizes the greyscale values in a black-to-white transition to determine the true edge. When digitizing the drop geometry using $\sigma = 0$ there resulting image is purely black and white and the edge detection will be more inaccurate compared to images with a narrow greyscale transition. For $\sigma = 0.5$ the edge detection algorithm returns sub-pixel locations that are very close to the real drop perimeter, thereby resulting in a very low fitting error.

All synthetic drops presented until now had realistic blurry edges but were otherwise noise free. In order to verify that our algorithm is able to produce correct results when including the noise from real images we have added Gaussian noise to the synthetic images. This is done by adding/subtracting Gaussian distributed numbers to all pixel values where the mean of the Gaussian is zero and the standard deviation is $\mu$ and pixel values range from 0 (black) to 1 (white). We have increased the standard deviation of the Gaussian noise from $\mu = 0$ (noise free) to $\mu = 0.12$ (noisy) and plotted the resulting error on the contact angle in Figure 3c).
Figure 3 Evaluation of absolute error on the contact angle estimation using synthetic images. (a) The absolute error in contact angle vs. the true contact angle from 10° to 170°. The curves results from fitting 500 drops with a one pixel gaussian distributed position of the drop centres. (b) The effect of varying the width of the black to white transition at the drop edge, simulating blurry images. (c) The effect of adding noise to the synthetic images.

In order to evaluate the performance of real images of drops we have captured drop profiles of seven sample drops shown in Figure 4. The structured surfaces for drops (3), (4), (6), and (7) in Figure 4., are: (3) randomly structured PP substrate, [14] (4) FDTS coated Si substrate fabricated by same method as in Søgard et al. [15] but with hierarchical pillar structures with 2% resulting surface coverage, (6) same
substrate as (3), and (7) same substrate as (4). These images are chosen to test a wide range of contact angles and some have been tilted in order to measure contact angle hysteresis. To get the range in contact angles the liquid and solid have been combined in the following way: 1) oleic acid on flat polypropylene, 2) water on flat ABS, 3) water on micro structured polypropylene, 4) water on micro and nano structured silicon surface coated with perfluorodecyltrichlorosilane (FDTS), 5-7) same as 2-4), tilted until onset of movement by 49.5°, 32.8° and 8.3° respectively. The drop in image 1, figure 4, does not start moving for any tilting value and is therefore not shown. For most of the drop images, there is excellent agreement between the contact angles calculated by elliptic and polynomial fitting, only for the drop on the micro and nano structured silicon surface there are significant deviations. For micro structured surfaces where the drop is resting on top of the asperities in the so called Cassie-Baxter state it has been shown that the real advancing contact angle is 180° due to the physical transition between the tops of asperities.[16] It is per se not problematic that the fitting algorithms undershoots this value since we know the true value but it shows the difficulties in determining very high contact angles. [17]

Figure 4 Calculated contact angles of 7 drops spanning from very low contact angles to very large, image 1-4 are resting on a horizontal surface while in image 5-7 the substrate have been tilted with the camera to obtain asymmetric drops. Except for contact angles above 160°, there is good agreement between the elliptic and polynomial fitting method.

We demonstrate the strength of the elliptic fitting method by measuring the dynamic contact angles in a tilting experiment. In this experiment we place a drop on the surface and tilt surface and camera until the drop slides or rolls off. During the experiment, the drop shape is captured by the camera and saved with the corresponding tilt value.

In our experimental setup the sample is placed on an x-y-z stage, which introduces slight mechanical instability, this makes the stage follow a slightly different trajectory than the camera. The different trajectory of stage and camera results in the image of the drop being shifted and rotated during the experiment. In order to subtract this mechanical shift we have recorded a calibration grid during a tilting
experiment. By tracing the calibration grid while tilting, we can obtain the shift and rotation of the stage in the camera view as a function of tilt angle. When analysing the frames captured during the experiment, the drop profiles are first extracted and then transformed to the coordinate system of the first frame by shift and rotation. This enables us to average baseline positions from all frames to determine the baseline with high accuracy. By having all drop profiles transformed to the same coordinate system, it is also possible to get precise information on the movement of the triple-line.

In Figure 5 we have plotted data obtained for a tilting experiment using a micro- and nanostructured silicon surface coated with FDTS with a 10 µl drop of 24% ethanol and 76% deionized water. The sample was tilted with 0.5°/s and captured with a framerate of 1 frame per second. With the information from the moving triple-line, we can see that the drop initially spreads by advancing on the left (downhill) side while being stuck on the right (uphill) side. At 10.8° tilt the right triple-line starts moving and the contact angles in this frame are taken as respectively the advancing and the receding contact angle. The slight negative movement of the right triple line just before the onset of movement is due to the optical distortions on the drop edge influencing the determination of the position of the triple-line. Without the information of the triple-line movement one could easily use the wrong frame resulting in erroneous roll-off and dynamic contact angles. After the onset of drop movement, it takes the drop several seconds with increasing inclination before the drop completely rolls off the surface. During this time there is a zipping like detachment process where the drop detaches individual pillars one a time.[18]

*Figure 5 Displacement of the right and left triple line together with the contact angles measured by polynomial and elliptic fitting during tilt experiment.*
During the tilting of the sample, we measure the same contact angle on the left side and a decreasing contact angle on the right side. This confirms that the contact angle hysteresis and the roll-off angle of drops in the Cassie-Baxter state are solely governed by the receding contact angle.[15]

![Figure 6: Fitting of the drop tilted 10° from the sequence presented in figure 5](image)

In the region just before drop movement starts, there is significant difference in the result between polynomial and elliptical fitting. The difference in determined contact angle arises due to lens effects in the drop producing white areas on the right side of the drop, especially close to the triple-line. An example of such optical distortion is shown in figure 6. Since the polynomial fitting is more sensitive to optical distortions it produces significant error in the contact angle whereas elliptical fitting is much more stable and shows a smooth decrease of the contact angle as a function of tilting angle. Since the lens effects often occur just before the onset of movement, it is crucial to use a method with the stability of our elliptic fit to measure the correct contact angle.

In conclusion, we have presented a new method for fitting and measuring contact angles by the tilting method. This we have done by fitting ellipses to left and right sides of the drop profile. The double-sided elliptical fitting method has been compared to the well-known polynomial fitting, and the implementation of both algorithms has been validated using realistic synthetic images. By using double sided elliptical fitting it is possible to achieve much higher tolerance for optical distortions of the drop profile. Finally, we have shown that this is crucial in tilting experiments where lens effects in the drop distort the receding side of the drop profile, particularly around the triple-line.
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