CTFFIND5 provides improved insight into quality, tilt and thickness of TEM samples

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Abstract

Images taken by transmission electron microscopes are usually affected by lens

aberrations and image defocus, among other factors. These distortions can be modeled in

reciprocal space using the contrast transfer function (CTF). Accurate estimation and

correction of the CTF is essential for restoring the high-resolution signal in an image and

has been one of the key aspects of the "resolution revolution" in cryogenic electron

microscopy (cryoEM). Previously, we described the implementation of algorithms for this

task in the *cis*TEM software package (Grant *et al.*, 2018). Here we show that taking sample

characteristics, such as thickness and tilt, into account can improve CTF estimation. This is

particularly important when imaging cellular samples, where measurement of sample

thickness and geometry derived from accurate modeling of the Thon ring pattern helps

judging the quality of the sample. This improved CTF estimation has been implemented in

CTFFIND5, a new version of the *cis*TEM program CTFFIND. We evaluated the accuracy of

these estimates using images of tilted aquaporin crystals and eukaryotic cells thinned by

focused ion beam milling. We estimate that with micrographs of sufficient quality

CTFFIND5 can measure sample tilt with an accuracy of 3° and sample thickness with an

accuracy of 5 nm.

Introduction

Transmission electron microscopy of biological specimens at cryogenic temperatures
(cryoEM) has become a widely used method to image biomolecules at high resolution, both
in solution and within the cell. To retrieve the high-resolution signal, the cryoEM images
have to be corrected for the contrast-transfer function (CTF) of the microscope. Common
parameters used to describe the CTF include an astigmatic defocus, the spherical
aberration of the objective lens, and if appropriate, a phase shift introduced by a phase
plate. These parameters are commonly estimated by fitting the Thon ring pattern (Thon,
1971) in the power spectrum of micrographs to a modeled power spectrum. The program
CTFFIND4 (Rohou & Grigorieff, 2015) has been developed for this task and the model and
conventions to describe the CTF are widely adopted in the field.

A limitation of CTFFIND4 is that it considers the whole imaged sample to be at the same
objective defocus, which is a reasonable assumption for flat and thin samples, as is common
in single-particle cryoEM. However, the increased thickness of cryoEM samples of cells may
introduce additional modulations in the Thon ring pattern (Tichelaar *et al.*, 2020) that can
lead to errors in the CTF modeling when not accounted for. Furthermore, samples of cells
are often tilted with respect to the optical axis of the microscope, either unintentionally due
to thinning methods such as cryogenic focused ion beam (FIB) milling, or intentionally
during electron cryo-tomography imaging. In both cases the effects are strongest at high-
resolution, where the Thon rings are more tightly spaced.

Here we describe new features of CTFFIND5 that can fit the modulations of the Thon ring
patterns and determine sample thickness and tilt using an extended CTF model with
additional parameters. This not only increases the fidelity of the fit, as Thon rings at higher
resolution can now be fitted reliably, but also gives valuable insight into the geometry of
the sample that can aid the experimentalist.

Methods

Tilt estimation algorithm

Tilt estimation in CTFFIND5 follows a strategy that is similar to the implementation in CTFTILT (Mindell & Grigorieff, 2003). The tilt axis direction ϕ and tilt angle θ are determined by fitting Thon ring patterns locally, calculated from 128 x 128 pixel tiles that form a regular grid covering the micrograph (Fig. 1a). In this model, ϕ has a positive value ranging from 0° to 360° to describe the angle to the X-axis of the micrograph. It is assumed that the defocus variation across the sample can be described by a tilted plane. Fits are evaluated using correlation coefficients between modeled CTFs and Thon ring patterns. Initially, the micrograph pixel size is adjusted (binned) by Fourier cropping to match the resolution limit of the fit set by the user and the micrograph is cropped to be square in order to speed up computation. A power spectrum is calculated from this binned and cropped image, a smooth background is calculated using a box convolution (Mindell & Grigorieff, 2003) and subtracted, the power spectrum is further binned to the tile size (128 x 128 pixels), and the fit of the tilted Thon ring patterns across the micrograph is initialized by fitting this highly binned power spectrum with a non-astigmatic CTF. This fit is then refined using a two-dimensional CTF with astigmatism. Rough values for the tilt axis and angle are then determined in a systematic search in 10° and 5°, respectively, using the locally fitted Thon ring patterns to score each pair of tilt axis and angle, followed by local refinement of tilt axis, angle and average defocus.

Finally, an average tilt-corrected power spectrum is calculated for diagnostic purposes and to allow the determination of a fit resolution. The tilt correction is designed to remove most of the Thon ring blurring due to the defocus variation across the image. To minimize ring blurring, the power spectrum from each tile is adjusted according to its local average defocus, $\Delta f_{average}$, by magnifying it by a factor m with

$$m = \sqrt{\Delta f_{local}/\Delta f_{average}} \qquad (1)$$

Since Δf_{local} will assume values across the image that are both smaller and larger than $\Delta f_{average}$, m will assume values smaller and larger than 1. The magnification / demagnification of the power spectrum compensates for the contraction / expansion of the Thon rings due to the local defocus change and produces approximately constant Thon ring patterns that can be averaged without losing the pattern (Fig. 1b). The compensation will have a small error if the spherical aberration is not zero. However, this error is sufficiently small to not visibly affect the Thon rings in the average.

Verification of tilt estimation using tilted aquaporin crystals

To test the robustness and accuracy of the new fitting algorithm, the defocus and sample tilts of aquaporin 2D crystals (Murata *et al.*, 2000) were estimated using a search range from 5000 Å to 50000 Å and a 100 Å step, low and high resolution limits of 30 Å to 5 Å, respectively, and a box size for the final power spectrum of 512 pixels. The estimated tilt angle θ and axis direction ϕ were compared with the values obtained by 2D crystallographic processing (Mindell & Grigorieff, 2003).

Verification of tilt estimation using tilt series

Lamellae prepared by FIB milling usually exhibit a pre-tilt with respect to the grid surface due to the stage tilt in the FIB instrument. In the microscope, the direction of this pre-tilt will generally not line up with the goniometer tilt axis. For the alignment of a tomogram recorded from such a lamella, the relative orientation of these two axes will have to be determined, together with the precise amount of pre-tilt. We wrote a new *cis*TEM (Grant *et al.*, 2018) program, called fit_tilt_model, to read the tilt angles and axes determined for each image in a tomographic tilt series and fit them to a model incorporating a pre-tilt and a single tomographic tilt axis. Using a rotation matrix R_0 to represent the pre-tilt and rotation matrices R_{tom}^i to represent the tomographic tilt angles and axis read from the microscope, the overall sample orientations are given by

106
$$R^i = R_0 \times R_{tom}^i \qquad (2)$$

R_0 and R_{tom}^i are calculated from the tilt angles θ and axes ϕ as

$$R = \begin{bmatrix} \cos(\phi)^2 + \sin(\phi)^2 \cos(\theta) & \cos(\phi)\sin(\phi)(\cos(\theta) - 1) & -\sin(\phi)\sin(\theta) \\ \cos(\phi)\sin(\phi)(\cos(\theta) - 1) & \cos(\phi)^2 \cos(\theta) + \sin(\phi)^2 & -\cos(\phi)\sin(\theta) \\ \sin(\phi)\sin(\theta) & \cos(\phi)\sin(\theta) & \cos(\theta) \end{bmatrix} \qquad (3)$$

In CTFFIND5, both tilt axis and angle are defined in the clockwise direction, with the angle
of the axis measured from the x-axis. This may be different from the definition used by the
microscope. To ensure consistency with the widely accepted angular convention in the
cryoEM field, all the θ and ϕ used in this manuscript refer to an anti-clockwise direction,
with ϕ measured from the x-axis.

Using the tilt information obtained with CTFFIND5, we now have a set of rotation matrices
R^i, and together with the rotation matrices read from the microscope, R_{tom}^i, we can
calculate a set of pre-tilt estimates R_0^i from equation (2). To determine the best overall pre-
tilt R_0, we determine the plane-normal vectors $V_{norm}^i = [x, y, z]$ of the sample by applying
R_0^i to the vector $[0,0,1]$ (z-coordinate along the beam direction), followed by calculating
their mean $V_{norm}^{mean} = [x_o, y_0, z_0]$ as the normal vector of the best overall pre-tilt estimate.
By calculating the root mean squared deviation of the normal vectors V_{norm}^i, outliers can be
identified and excluded to further refine V_{norm}^{mean}. The pre-tilt can then be determined as:

125
$$\theta_0 = \begin{cases} \cos^{-1}(z_0) & x_0 \geq 0 \\ -\cos^{-1}(z_0) & x_0 < 0 \end{cases}$$
$$\phi_0 = \begin{cases} \tan^{-1}(\dfrac{x_0}{y_0}) & y_0 \neq 0, \phi_0 \in [0,180] \\ 90° & y_0 = 0 \end{cases} \qquad (4)$$

To generate more reliable defocus and tilt estimates, the defocus search range and

resolution fitting range can be adjusted according to the experimental tilt range and image

quality. For our cryoEM samples, the low and high resolution limits were set to 50 Å to 10

Å, respectively, and the defocus search interval was set to be between $\pm$ 10000 and

± 20000 Å from the nominal defocus set during data collection.

Sample thickness estimation

In CTFFIND5 we implemented a new CTF_t model function, based on the CTF function

implemented in CTFFIND4 (Rohou & Grigorieff, 2015) and extended by the formula

described by (McMullan *et al.*, 2015):

$$CTF_t(\lambda, \mathrm{g}, \Delta f, C_s, \Delta\varphi, \omega_2, t) = \frac{1}{2}\left(1 - \mathrm{sinc}\big(\xi(\lambda, \mathrm{g}, t)\big)\cos\big(2\chi(\lambda, |\mathrm{g}|, \Delta f, C_s, \Delta\varphi, \omega_2)\big)\right) \qquad (5)$$

where χ denotes the phase-shift as a function of the electron wavelength λ, the spatial

frequency vector $|\mathrm{g}|$, the objective defocus Δf, the spherical aberration C_s, the additional

phase shift $\Delta\varphi$, and the fraction of amplitude contrast ω_2. The modulation of the CTF due to

sample thickness t is described by the function ξ:

$$\xi(\lambda, \mathrm{g}, t) = \pi\lambda\mathrm{g}^2 t \qquad (6)$$

If a user requests sample thickness estimation, the program will first fit the CTF model

function as implemented in CTFFIND4 and the "goodness of fit" resolution will be used as

an estimate of the frequency g of the first node of the CTF_t function, with t given by:

$$t = \frac{1}{\lambda\mathrm{g}^2} \qquad (7)$$

If the option "Brute-force 1D fit" is selected, CTFFIND5 will further refine t and Δf by

calculating the normalized cross-correlation between the radial average of the power

spectrum (corrected for astigmatism, as described in) and CTF_t, searching systematically

for the best combination of t in the range of 50-400 nm in 10 nm steps, and Δf in the range

of ± 200 nm from the previously fitted value, also in 10 nm steps.

Finally, if the option "2D-refinement" is selected, CTFFIND5 will optimize t, Δf_1, Δf_2, and ω

using the same conjugate gradient algorithm used in CTFFIND4 and the normalized cross

correlation between CTF_t and the 2D power spectrum as a scoring function.

After the optimal values for t and Δf have been obtained the "goodness of fit"

crosscorrelation is recalculated using CTF_t, with a frequency window that is 1.5 time larger

than in CTFFIND4 to avoid the drop-off in the node regions of CTF_t.

Verification of sample thickness estimation using Lambert-Beer's law

We used 655 micrographs collected from one lamella of ER-HoxB8 cells (dataset

Lamella$_{EUC}$1 from (Elferich *et al.*, 2022)). For each micrograph we calculated $ln\left(\frac{I}{I_0}\right)$, where

I was the sum of all pixels in the illuminated area of the movie and I_0 was the average of

this sum for 45 micrographs collected over vacuum with the same energy filter settings.

This value is expected to have a linear relationship with the thickness of the sample

consistent with Lambert-Beer's law (Yan et al. 2015; Rice et al. 2018):

$$ln\left(\frac{I}{I_0}\right) = \frac{1}{\kappa}t \qquad (8)$$

where κ is the apparent mean free path for inelastic scattering.

We then used CTFFIND5 to estimate the thickness t of each micrograph using the "Brute-

force 1D fit" and "2D-refinement" setting, low and high resolution limits set to 30 Å and 5 Å,

defocus search range set between 500 nm and 5000 nm, and low and high resolution limits

for thickness estimation set to 10 Å and 3 Å. We used a "RANSAC" algorithm as

implemented by the scikit-learn Python package (Pedregosa *et al.*, 2011) to fit a linear

model to the relationship of $ln\left(\frac{I}{I_0}\right)$ and t, while rejecting outliers. We then manually

inspected every outlier of the model fit and categorized the reason for the discrepancy into

"Occluded beam" (either from contamination or the edges of the lamella), "Low image

signal" (in most cases exposures containing no cellular features), "Carbon/Platinum", and

"Lipid droplet" (see Fig. 4).

Verification of sample thickness estimation using tomography

Lamellae prepared from ER-HoxB8 cells were imaged using a Titan Krios 300 keV TEM

controlled by SerialEM (Mastronarde, 2005). For each dataset an initial exposure was taken

with a magnification of 64,000, resulting in a pixel size of 1.6 Å and an exposure of 30 e⁻/Å.

This was followed by the acquisition of a tilt series at a magnification of 48,000, resulting in

a pixel size of 2.087 Å. A total of 35 tilt images at a tilt interval of 3° were collected from -

51° to 51°, relative to the milling angle, using a grouped dose-symmetric scheme (Hagen *et*

al., 2017). The exposure per tilt was 3 e⁻/Å, resulting in a total exposure of 105 e⁻/Å .

For tomographic reconstruction, tilt movie frame motion correction was performed using

SerialEM (Mastronarde, 2005), and tilt series were aligned using the IMOD software

package (version 4.11, Mastronarde & Held, 2017). For coarse alignment, a high-frequency

cutoff radius of 0.15 was used. A fiducial model was generated using patch tracking with

patches of 450 x 450 pixels and a fractional overlap of patches of 0.33 x 0.33. High-tilt

frames were omitted while generating the fiducial model. Robust fitting with a tuning

factor of 1 was used for fine alignment. After computing the alignment, the fiducial model

was edited by removing unreliable patches, and then alignments were re-computed. The

edited models with the lowest residual mean errors and standard deviations were used for

fine alignment. Tomogram positioning was used to correct the tilt angle offset. Fully aligned

stacks were generated with a binning factor of 4, resulting in a tomogram pixel size of 8.3 Å.

Tomograms were reconstructed using the SIRT-like filtering option in IMOD (Mastronarde

& Held, 2017; Mastronarde, 1997) and manually inspected. The tomograms were back-

projected along the y-axis using a homemade script, generating a small set of XZ

projections. Thickness measurements on the projected central slides were performed using

the display program included with the *cis*TEM software package (Grant *et al.*, 2018).

CTF correction of medium magnification lamella images

The CTF of the representative medium magnification image with a pixel size of 40 Å was

estimated using CTFFIND5 with the following parameters: defocus range: 1,000,000 to

4,000,000 Å; search step 50,000 Å; low and high resolution limits: 400 Å and 80 Å. We then

used the program apply_ctf, included with *cis*TEM, to flip the phases according to the

estimated CTF. We furthermore implemented the Wiener like filter described in (Tegunov

& Cramer, 2019) in apply_ctf to produce the image shown in Fig. 6d.

Benchmarking CTFFIND5 runtimes

CTFFIND5 runtimes were measured using 3 representative micrographs (Table 2). As a

baseline measurement, CTFFIND5 was run without estimation of tilt and sample thickness

enabled. Then runtime was measured enabling either one of these option or both. Every

test was repeated four times and the average and standard deviation of the last three runs

are reported, to minimize the contribution of hard-drive speed. The tests were performed

on a single core of an Intel Core i9-12900KF CPU.

Results

Tilt estimation

We tested the defocus correction for the Thon rings on a representative micrograph taken

from a cryo-FIB milled lamella. As expected, the correction results in the observation of

Thon rings at higher spatial resolution (Fig. 1c). In this example, correcting for the

estimated moderate tilt of 12.3° improved the highest resolution at which a reasonable fit

could be obtained from 5.9 Å to 4.6 Å. The power spectrum also appears less noisy, which

can be attributed to some low-pass filtering that occurs with the interpolation of the Thon

ring patterns of individual tiles to perform the defocus correction.

To test the performance of the new CTFFIND5 sample tilt estimation, we used a dataset of

images of tilted aquaporin crystals that were also used to benchmark the original CTFTILT

implementation (Mindell & Grigorieff, 2003; Murata *et al.*, 2000). Table 1 compares the tilt

information of the samples obtained from crystallographic analysis and the estimates

obtained using CTFFIND5. Overall, the results of CTFFIND5 agree well with the aquaporin

crystals information. The average discrepancy was 1.9° for the tilt axis direction and 1.5°

for the tilt angle.

To test whether CTFFIND5 would be able to correctly assign tilt axis and angle for tilt

series data, we analyzed two tilt series from different grids of lamellae prepared by cryo-

FIB milling from mouse neutrophil-like cells (Elferich *et al.*, 2022). We then plotted the

estimated values for tilt axis and angle as a function of nominal stage tilt (Fig. 2). The

estimated tilt angle shows a roughly linear relationship with the nominal stage tilt, but

since CTFFIND5 reports only positive tilt angles the overall plot has a chevron-shape. The

estimated tilt axis angle is approximately constant at high tilts but changes by about 180° at

0° estimated tilt, again due to the convention enforced by CTFFIND5. Notably, in both

examples there is an offset of about 20° between nominal and estimated tilts, which is due

to the pre-tilt of the specimen caused by FIB-milling at a shallow angle. To quantify and

delineate both the tilt axis direction of the microscope and the pre-tilt of the specimen we

fit all values to a model as described in Methods (Fig. 2). The fitting resulted in an

estimated tilt axis angle of 178.2° and 179.8°, respectively, which is consistent with the

SerialEM calibration of 178.4° and 176.3° for the stage tilt axis. The estimated pre-tilt

values were 20.6 °and -21.9°, consistent with a FIB-milling angle of 20° and opposite

orientation of the grids relative to the milling direction. The pre-tilt axis angles were

estimated as 171.8° and 183.8°, which is consistent with the error expected from manually

aligning the milling direction when inserting grids into the microscope.

To estimate the accuracy of the tilt estimation in tilt series, we calculated the mean absolute

difference between the tilt and axis-angle estimates and the fitted model, excluding the axis-

angle estimates at tilt angles under 5°. For the first tilt series we obtained accuracy estimates

of 2.08° and 2.58° for tilt and axis-angles, respectively. In the second tilt series the accuracy

estimates were 3.95° and 9.47°. In both cases the accuracy was lower than for the tilted

aquaporin crystals, presumably due to the relatively short exposure of each micrograph in

the tilt series. However, the substantially higher mean differences in the second tilt series

suggest that the accuracy is highly dependent on the quality of the underlying data.

Sample thickness estimation

Even after correcting for sample tilt we found that for FIB-milled samples we often could

observe Thon-ring like modulation in the power spectrum at higher resolution than

suggested by the goodness of fit estimate (Fig. 3b, top plot). These modulations are out of

phase with the predicted modulations, as described by (McMullan *et al.*, 2015) and

(Tichelaar *et al.*, 2020). We therefore implemented an extension of the CTF model as

described by (McMullan *et al.*, 2015) (Fig. 3a). For some images we found that the thickness

could be well estimated by assuming that the goodness of fit resolution estimate obtained

using the old model implemented in CTFFIND4 corresponds to the first node in the

modulation function, according to Eq. (7). With our new model, estimated CTF parameters

were very similar to those from CTFFIND4, but the fit in CTFFIND5 extended to higher

resolution (Fig. 3b).

In other images, mostly with defocus values under 1 μm and with a sample thickness over

200 nm, CTFFIND4 could fit the power spectrum before and after the first node using the

old CTF model, with some deviations between the fit and the power spectrum (Fig. 3c).

Fitting the power spectrum with the new model in CTFFIND5 resulted in substantially

different estimated CTF parameters and an improved fit, even though the goodness-of-fit

estimation did not change. Based on these results we conclude that CTFFIND5 will provide

more accurate CTF parameters for images of thick samples, such as those generated from

FIB-milling. In addition, the fit provides a direct readout of the specimen thickness, which is

important for judging specimen quality and the potential for high-resolution information

that can be recovered from these images.

Estimating the accuracy of sample thickness estimation using the Lambert-Beer law on energy filtered data

CryoEM is frequently performed using an energy filter to remove inelastically scattered electrons. The fraction of inelastically scattered electrons can be described by the Lambert-Beer law, which states that the fraction of electrons removed from the image is proportional to the thickness of the sample. The apparent mean free path for electron scattering has been experimentally determined for common cryoEM conditions (Rice *et al.*, 2018). To test whether thickness estimation in CTFFIND5 is consistent with this method we used a dataset of 655 exposures of a lamella of ER-HoxB8 cells collected using the DeCo-LACE approach (Elferich *et al.*, 2022). We used CTFFIND5 to estimate the thickness t of every exposure and plotted $-ln\left(\frac{I}{I_0}\right)$ against t (Fig. 4). Fitting the data to a linear model described in Methods (Eq. (8)), we found that 568 out of 655 exposures followed closely a linear relationship with a mean free path κ of 317 nm. Manual inspection of images that did not follow this linear relationship revealed that they either contained visible ice contamination, platinum deposits, or they were collected over ice without cellular features and displayed weak Thon rings. The value of κ is consistent the value found by (Rice *et al.*, 2018), even though our dataset was collected without an objective aperture. The x-axis intercept of the linear model was -14.1 nm, meaning that the node position systematically predicts a smaller thickness than predicted by the Lambert-Beer law. This discrepancy is further discussed in the next section. To estimate the accuracy of the sample thickness determined by CTFFIND5 we calculated the mean absolute difference to the linear model, which was 4.8 nm. These data suggest that sample thickness determination using node-fitting is an alternative to using Lambert-Beers law that has the advantage of not relying on the constant κ and the intensity I_0, both of which might not be readily available. Also, the two approaches are complementary as they rely on orthogonal mechanisms.

Estimating the accuracy of sample thickness estimation using tomography

We used a dataset of seven micrographs collected from lamellae of ER-HoxB8 cells together with tilt series collected afterwards from the same locations to verify the accuracy of the

thickness estimates obtained using CTFFIND5. We used CTFFIND5 to estimate the

thickness ($t_{CTFFIND}$) for every location and compared it with the thickness estimated from

the tomogram reconstructed from the tilt series (t_{TOMO}). We measured t_{TOMO} by manually

estimating the distance between the surfaces of the lamella in three different positions.

When we plotted $t_{CTFFIND}$ against t_{TOMO} we found that the values were highly correlated,

but t_{TOMO} was consistently smaller than $t_{CTFFIND}$ (Fig. 5). A linear fit revealed a slope of

0.95 and a y-axis intercept of 0.12 nm. This means that the CTFFIND5 thickness estimate is

on average 1.05x higher than the thickness estimated by tomography. (Tichelaar *et al.*,

2020) also report that estimating the thickness from the CTF nodes resulted in values

roughly 1.1x higher than estimated by tomography. The reasons for the systematic

discrepancies between thicknesses estimated by CTFFIND5 and estimates based on

Lambert-Beer's law and tomography are unclear, but since they are small and CTFFIND5

estimates lie in between the other two estimates, they will provide comparable

information.

CTF estimation and correction assists biological interpretation of intermediate-magnification

lamella images

During data collection of cryoEM data in cells, the operator frequently relies on images

taken at low magnification to select areas of interest and establish their biological context.

The pixel size of these images is usually about 40 Å, with a defocus of about 200 μm. This

produces strong contrast from biological membranes, but can sometimes also lead to

substantial fringes near these membranes (Fig. 6a). We found that a simple CTF correction

based on CTFFIND defocus estimates obtained from the overview images can reduce these

fringes (Fig. 6b). A simple CTF correction can be done using the program apply_ctf,

included with *cis*TEM, by phase flipping according to the fitted CTF (Fig. 6c). However, we

found that including a Wiener filter-based amplitude correction describe by (Tegunov &

Cramer, 2019) produces a more naturally looking image that might be best suited to

recognize cellular features (Fig. 6d).

CTFFIND5 runtimes

To gauge the ability of CTFFIND5 to provide real time feedback during cryoEM data collection we measured its runtime on three representative micrographs (Table 2). Without estimation of tilt or sample thickness CTFIND5 performed CTF estimation roughly within a second. Estimation of the sample thickness adds roughly half a second to the runtime, therefore allowing CTF estimation within a timeframe comparable to typical exposure times. Estimation of the tilt on the other hand increased runtimes substantially to the order of several minutes, due to the exhaustive search of potential tilts over hundreds of powerspectra. While these runtimes are substantially slower than cryoEM data acquisition, near real time estimation can be achieved by using multiple CPU cores. Furthermore, optimization of the number of tiles used, better search algorithms, or implementations employing GPUs could increase the speed to the point where real time estimation is more feasible.

Conclusion

The new features implemented in CTFFIND5 improve CTF estimation from the power spectra of cryoEM micrographs where assumptions made in its predecessor, CTFFIND4, namely a thin and untilted sample, do not hold. The tilt of the sample is estimated by fitting the CTF to the power spectra calculated from small patches across the image, similar to other software including CTFTilt (Mindell & Grigorieff, 2003), Ctfplotter (Xiong *et al.*, 2009; Mastronarde, 2024), goCTF (Su, 2019), and Warp (Tegunov & Cramer, 2019). After estimation of the sample tilt a tilt-corrected power spectrum is produced that exhibits stronger Thon rings at higher resolution.

To take into account the modulation of the power spectra by thick samples (Tichelaar *et al.*, 2020; McMullan *et al.*, 2015) we fit a modified CTF model, which increases the resolution of the fitted regions of the spectra and provides a read-out of the sample thickness. While the low exposures (3-5 e$^-$/A2) typically used in electron cryo-tomography often preclude fitting of sample thickness from power spectra of individual images in the tilt series, we demonstrate that this works reliably for higher exposures ($\sim$ 30 e$^-$/A2) typically used for

2D template matching (Lucas *et al.*; Rickgauer *et al.*, 2017) and in-situ single particle

analysis (Cheng *et al.*, 2023).

While these improvements are especially relevant for in-situ samples, e.g., prepared by

cryo-FIB milling, the analysis of images of purified samples recorded at lower acceleration

voltages, e.g., 100 keV (McMullan *et al.*, 2023), may also benefit since thickness-dependent

CTF modulations will appear at lower resolution with longer electron wavelengths (see Eq.

(6)). Per-micrograph CTF estimation can be followed by per-particle CTF refinement, as

implemented in *cis*TEM (Grant *et al.*, 2018), Relion (Kimanius *et al.*, 2021), or cryoSPARC

(Punjani *et al.*, 2017). The improvements of CTFFIND5 will provide better starting values

for this refinement, yielding better overall CTF estimation and recovery of high-resolution

information during 3D reconstruction.

In summary, the improvements implemented in CTFFIND5 result in more accurate CTF

estimation of thick and tilted samples and provide valuable information about the samples

to the microscopist.

Data availability

The images of tilted aquaporin crystals were previously published (Murata *et al.*, 2000) and

are available at https://grigoriefflab.umassmed.edu/tilted_aquaporin_crystals . The

untilted exposures of ER-HOXB8 cells are available at EMPIAR (EMPIAR-11063). The

tomograms and tilt series from ER-HOXB8 cells have been deposited to EMDB (EMD-

43419, EMD-43420, EMD-43424, EMD-43425, EMD-43427, EMD-43428, EMD-43429) and

EMPIAR (EMPIAR-11854), respectively. The source code for CTFFIND5 is available at

https://github.com/GrigorieffLab/cisTEM/tree/ctffind5 and binaries for most Linux

distributions can be downloaded at https://cistem.org/development .

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Tables

406   **Table 1:** Comparison of CTFFIND5 estimation of sample tilt with crystallographic analysis

Image	Axis angle φ			Tilt angle θ		
	crystallog.	ctffind5	$\Delta\varphi$	crystallog.	ctffind5	$\Delta\theta$
530394	93.28	94.98	-1.7	19.6	20.69	-1.09
530419	109.78	106.51	3.27	18.66	16.04	2.62
530430	104.38	101.13	3.25	21.32	20.37	0.95
530444	98.39	97.62	0.77	20.72	20.88	-0.16
660027	99.68	102.34	-2.66	19.4	22.39	-2.99
540149	94.45	85.84	8.61	43.08	44.59	-1.51
540291	96.16	98.1	-1.94	45.11	40.68	4.43
540302	93.98	93.39	0.59	44.7	44.21	0.49
540313	95.34	95.13	0.21	44.03	46.49	-2.46
660183	97.69	97.27	0.42	48.13	48.99	-0.86
550069	90.08	92.55	-2.47	60.46	60.83	-0.37
550089	91.48	92.04	-0.56	60.5	60.72	-0.22
660291	93.23	92.19	1.04	57.59	59.19	-1.60
660421	89.32	89.06	0.26	61.36	60.01	1.35

680341	89.67	90.02	-0.35	58.68	59.62	-0.94
530345	N/A	108.6		0	0.84	-0.84
530356	N/A	231.17		0	1.93	-1.93
530358	N/A	56.58		0	1.29	-1.29
530375	N/A	3.21		0	0.79	-0.79
530378	N/A	67.6		0	2.17	-2.17

409   **Table 2:** Runtime of CTFFIND5 on representative micrographs

Micrograph		1	2	3
Image properties				
Image size		4070x2892	2880x2046	4746x3370
Pixel size (Å)		1.5	4.175	2.5
Runtime (s)				
Tilt	Thickness			
-	-	0.9±0.1	0.7±0.1	1.7±0.1
+	-	39.0±0.2	208±1	173.4±0.1
-	+	1.4±0.1	1.3±0.1	2.4±0.1
+	+	39.5±0.1	209±1	173.0±0.1

Figures

Figure 1: Tilt estimation and correction in CTFFIND5. (a) Power spectra are calculated in
128x128 pixel patches as indicated on a representative micrograph. The dots represent the
locations of the patches and the boxes indicate patch size. (b) A model of the expected
power spectrum in each patch given an average defocus Δf, tilt angle θ, and tilt axis ϕ is
compared to the actual power spectra of tiles. After an optimal set of θ and ϕ has been
found a corrected power spectrum is calculated by summing the tile power spectra, scaled
to correct for the defocus difference. (c) Comparison of the original power spectrum (solid
line, blue) to the tilt-corrected power spectrum (solid line, black). The tilt-corrected power
spectrum exhibits clear peaks at higher spatial resolution than the uncorrected power
spectrum, as evident by the "goodness-of-fit" scores (dashed lines). The estimated CTF

424   parameters are $\Delta f_1 = 10603$ Å, $\Delta f_2 = 10193$ Å, $\alpha = 85.9°$ for the uncorrected power

425   spectrum and $\Delta f_1 = 10492$ Å, $\Delta f_2 = 10342$ Å, $\alpha = 81.2°, \theta = 12.3°, \phi = 261.6°$ for the tilt-

426   corrected power spectrum. The fit resolution is 5.9 Å for the uncorrected power spectrum

427   (dashed line, blue) and 4.6 Å for the tilt-corrected spectrum (dashed line, black).

Figure 2: Validation of tilt estimation using tilt series data. (a) Estimated tilt angle and axis
of 40 micrographs of a tilt series taken on a FIB-milled biological specimen. For each image
the tilt angle (dots, upper plot) and tilt axis direction (crosses, middle plot) are plotted as a
function of the nominal stage angle. The data were fitted to a model of the specimen tilt and
constant stage tilt axis before tilting the stage. The estimated stage tilt axis has an angle of
171.8° and the estimated specimen pre-tilt is 20.6° with a tilt axis of 171.8°, which is
consistent with the FIB-milling angle of 20° and manual alignment of the milling direction
to the goniometer tilt axis. In the bottom plot the fit residuals for tilt angle and axis are

438   plotted. (b) Data for another tilt series plotted as described for (a). The estimated stage tilt

439   axis is 179.8°, the estimated specimen pre-tilt is -21.9° with a tilt axis of 183.8°. This is

440   consistent with this grid being inserted in the opposite orientation as the grid shown in (a),

441   but still with a rough alignment of milling direction and tilt axis.

Figure 3: Sample thickness estimation by fitting Thon ring patterns. (a) Comparison of the
CTF model used in CTFFIND4, and after applying the modulation function (right) described
by (McMullan *et al.*, 2015). (b) Representative example of Thon ring fitting in a lamella
without (top) and with (bottom) thickness estimation. The tilt of the specimen was

estimated to be 12.3°. When fitting without thickness estimation the estimated parameters

were $\Delta f_1 = 10492$ Å, $\Delta f_2 = 10342$ Å, $\alpha = 81.2°$. When taking sample thickness into account

the estimated parameters were $\Delta f_1 = 10481$ Å, $\Delta f_2 = 10286$ Å, $\alpha = 69.6°, t = 969$ Å. The

estimated fit resolution was 4.6 Å and 3.4 Å without and with sample estimation,

respectively. (c) Representative example of Thon ring fitting in a lamella without (top) and

with (bottom) thickness estimation. The tilt of the specimen was estimated to be 6.7°.

When fitting without thickness estimation the estimated parameters were $\Delta f_1 =$

8002 Å, $\Delta f_2 = 7717$ Å, $\alpha = 73.4°$. When taking sample thickness into account the estimated

parameters were $\Delta f_1 = 8549$ Å, $\Delta f_2 = 8343$ Å, $\alpha = 63.3°, t = 2017$ Å. The estimated fit

resolution was 4.3 Å and 4.2 Å without and with sample estimation, respectively.

Figure 4: Validation of sample thickness estimation in CTFFIND5 by comparing the estimates to the intensity attenuation by the zero-loss energy filter. An estimation of the linear relationship using the RANSAC algorithm results in a slope of 1/316.6 nm and an x-axis intercept at -14 nm (red dashed line). Data points that were labeled as outliers by the RANSAC algorithm were manually inspected and color-coded according to visual inspection of the micrographs.

468   **Figure 5:** Validation of sample thickness estimation in CTFFIND5 by tomography. The
469   distribution of thickness measurements in seven tomograms are shown as box plots with
470   the median indicated by a red line. The position on the x-axis corresponds to the thickness
471   estimate by CTFFIND5. The black dashed line indicates identity.

Defocus: 222.7 μm, Fit resolution: 67.6 Å

473   **Figure 6:** CTF correction of medium magnification overviews. (a) Representative area of a

474   micrograph of a cellular sample at a pixel size of 40 Å without CTF correction. (b) Fit of the

475   power spectrum of the micrograph shown in panel a CTF model. (c-d) The same

476   micrograph as shown in panel (a) after CTF correction by phase flipping (c) or with a

477   Wiener-like filter (d)

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