



charges (e.g., screening clouds), and polar-

prism is oriented along the  $y$  axis, then the hologram of an ideal imaging system can be expressed in the form

$$\left| \psi\left(x - \frac{d}{2}, y\right) \exp\left(\frac{i\pi x}{s}\right) + \psi\left(x + \frac{d}{2}, y\right) \exp\left(-\frac{i\pi x}{s}\right) \right|^2, \quad (2)$$

where  $d$  is the interference distance (which depends on the biprism geometry) and  $s$  is the spacing of the interference fringes in a vacuum hologram recorded with the specimen removed from the field of view (FOV). Eq. (2) describes two twin images of the object wave-

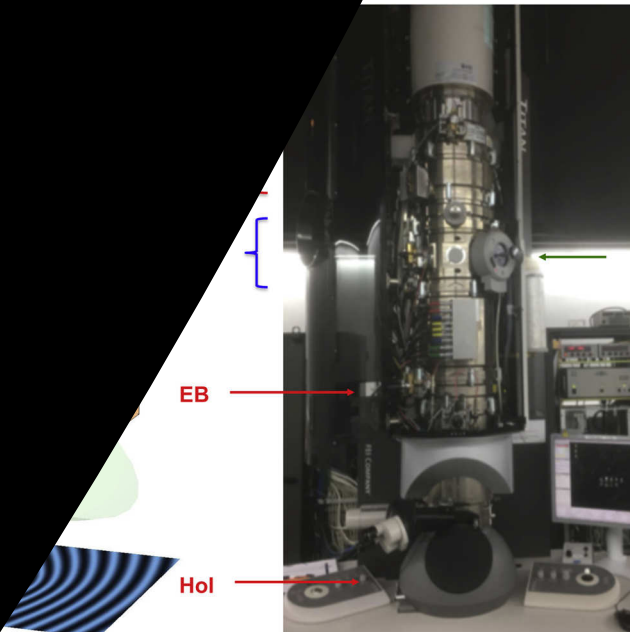
function. If the object (specimen) is located at  $x + d/2$ , then the corresponding object wave  $\psi\left(x + \frac{d}{2}, y\right)$  is overlapped with the reference wave  $\psi\left(x - \frac{d}{2}, y\right)$ . In order to retrieve the amplitude and phase of the object wave, the reference wave should ideally be equal to unity, or it should be known.

When long-range electrostatic fields originate from the specimen, as shown in Fig. 1, the reference wave may be perturbed. Analysis of the hologram then results in the reconstruction of a fictitious specimen, which can be described by the wavefunction [8,9]

$$\psi(x, y) = a(x, y) \exp \left[ i\varphi\left(x + \frac{d}{2}, y\right) - i\varphi\left(x - \frac{d}{2}, y\right) \right], \quad (3)$$

where  $a(x, y)$  is the amplitude of the object wave,  $d$  is a two-dimensional “interference distance” vector that connects the two virtual sources created by the biprism, and  $\varphi\left(x + \frac{d}{2}, y\right)$  and  $\varphi\left(x - \frac{d}{2}, y\right)$  are the phases of the object and reference waves, respectively. The difference between these two phase distributions, rather than the true object phase, is then recovered. The influence of such a perturbed reference wave (PRW) on measurements of charge density and electric field is

**Fig. 1.** Schematic diagram of the experimental setup for off-axis electron holography (left) and photograph of an FEI Titan transmission electron microscope in Forschungszentrum Jülich (right). Corresponding components are labelled using the same colours. From top to bottom are: an illuminating plane or spherical electron wave  $I$ , an electrically-biased needle-shaped specimen  $Sp$ , the electron microscope objective lens  $OL$ , an electron biprism  $EB$ , and a recorded off-axis electron hologram  $Hol$ .



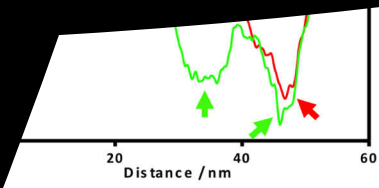


Figure 2. (a) Phase image of a needle-shaped W specimen covered by an amorphous layer: (a) recorded using off-axis electron holography. (b) Charge density distribution calculated from the phase image using Eq. (7) on the vacuum side of the specimen edge, while an effective band of charge is present at the interface between the needle and the amorphous layer. The Laplacian of the phase across regions A (green) and B (red) marked in (b) is calculated using Eq. (7) applied to a median-filtered version of the phase. See text for details.

where  $\rho_{\text{proj}}(x, y) = \int \rho\left(x - \frac{d}{2}, y, z\right) dz$  is the projected charge density and  $Q_C$  is the total charge present in region  $C$ . The volume over which Gauss' law is evaluated, as discussed above, is an infinite cylinder (along the  $z$  axis), of which  $C$  is a cross-section.

Furthermore, the Laplacian of the phase can be calculated directly from the reconstructed complex wavefunction using the expression [25]

$$\nabla^2 \varphi = \text{Im} \left[ \frac{\nabla^2 \psi}{\psi} - \left( \frac{\nabla \psi}{\psi} \right)^2 \right]. \quad (8)$$

By making use of the divergence theorem, Eq. (7) can equivalently be written in the form [17]

$$Q_C = -\frac{\epsilon_0}{C_E} \oint_{\partial C} \nabla \varphi_{\text{rec}}(x(l), y(l)) \cdot \mathbf{n}(x(l), y(l)) dl, \quad (9)$$

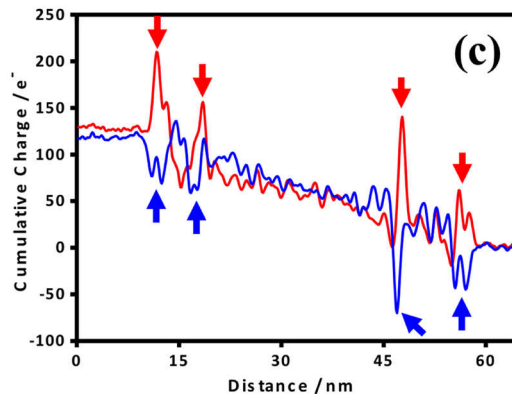
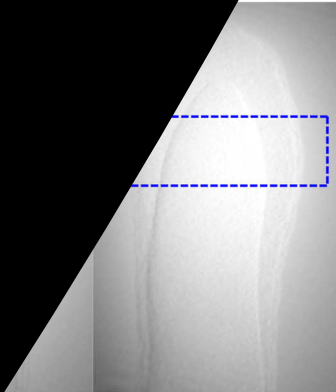
where  $\nabla$  is a two-dimensional gradient operator,  $\partial C$  denotes an integration loop (coinciding with the boundary of the integration region  $C$  in Eq. (7)),  $l$  is a curvilinear coordinate along the contour and  $\mathbf{n}$  is the outward normal to the contour.

## 2.2. Practical considerations

Parameters that can affect charge density measurements include the MIP contribution to the phase, the spatial resolution (*i.e.*, the digital undersampling) of the recorded phase image, its signal-to-noise ratio (SNR), strong diffraction conditions (which can affect the measurement of the MIP contribution to the phase), electron-beam-induced specimen charging effects and the influence of sample imperfections (*e.g.*, damage, contamination, and oxidation). Several of these considerations are now discussed.

### 2.2.1. Mean inner potential

The MIP of the specimen affects charge density measurements from electron optical phase images because it is associated with the presence of effective local dipole layers at the specimen surface [42]. Its influence is illustrated in Fig. 2. Fig. 2a shows part of a phase image of the end of an unbiased W needle recorded using off-axis electron holography. The needle is surrounded at its end by a layer of amorphous oxide and/or contamination. Fig. 2b shows the projected charge density distribution calculated directly from the Laplacian of the recorded phase image using Eq. (7). Evaluation of the Laplacian of the phase invariably results in a noisy image. In addition, Fig. 2b reveals that local variations in specimen thickness and MIP are visible in the form of



**Fig. 3.** Charge density distributions calculated by applying a loop integral (evaluated using Eq. (9)) to differences between phase images recorded with and without an electrical bias voltage applied to the W needle shown in Fig. 2. The phase image recorded with a bias voltage applied to the needle was deliberately misaligned by +5 and -5 pixels along the x axis in (a) and (b), respectively, with respect to the phase image recorded without an applied bias, before evaluating their difference. (c) Cumulative charge profiles obtained by integrating the signal across the regions marked in (a) and (b).

ratio  
charge measurement is determined by factors that the original hologram, the sampling density of the the size of the integration region in the loop integral experimental phase image (neglecting scattering ab specimen) is considered to be a superposition of a noise- phase image and random normally-distributed noise with and standard deviation (SD)  $\delta\phi$ , then the Laplacian of the can also be regarded as a noise-free charge distribution . Here, we show how the SD of the measured charge density on  $\delta\sigma$  is related to the SD of the phase.

The discrete Laplacian is a one-step matrix algebra operator maps each pixel in a phase image  $\phi(i, j)$  onto the value  $\phi(i+1, j) + \phi(i-1, j) + \phi(i, j+1) + \phi(i, j-1) - 4\phi(i, j)$ . If this operation is applied to a noisy phase image that has zero mean and  $\delta\phi(i, j)$ , then the result is another noisy image, which is also normally distributed and has zero mean (because  $1 + 1 + 1 + 1 - 4 = 0$ ), but which has a standard deviation that is  $\sqrt{20}$  times larger than the SD of the original image. This description is valid when each noise pixel is uncorrelated with its neighbours and when the SDs can be added in quadrature, such that  $1^2 + 1^2 + 1^2 + 1^2 + 4^2 = 20$ . The relationship between the SDs of the phase and the charge density is given by the expression

$$\delta\sigma = \sqrt{20} \frac{\epsilon_0}{C_E} \frac{\delta\phi}{p^2} = \sqrt{20} \frac{\delta q}{p^2}, \quad (10)$$

where  $p$  is the pixel size, and we define  $\delta q = \frac{\epsilon_0}{C_E} \delta\phi$  as the charge noise. For reference,  $\delta\phi = 118 \text{ mrad}$  corresponds to  $\delta q = 1e$  at 300 kV. Since experimental values of phase noise SD are typically well below 100 mrad, the achievement of single electron sensitivity in charge measurement appears to be relatively straightforward.

Eq. (10) is derived on the assumption of uncorrelated/white noise in the phase image. However, this situation does not strictly hold for Fourier-transform-based hologram reconstruction, as noise correlations are automatically introduced when a side-band is masked using an aperture. (A general description of the introduction of correlation in the reconstruction of holograms can be found elsewhere [44].) Such a mask may be “soft” (e.g., Gaussian, Hann, or Butterworth), or “hard” (e.g., top hat). Soft apertures are most commonly used because they are more efficient than hard apertures at suppressing phase noise without

$$= \begin{pmatrix} 0 & \dots & 0 \\ \vdots & L_{22} & \dots & L_{26} & \vdots \\ 0 & L_{62} & \dots & L_{66} & 0 \end{pmatrix}$$

Fig. 4. (a) Simulated correlation matrices. (c) Experimental noise image, (d) Corresponding experimental correlation matrix. Right: (e) Experimental correlation matrix.

Experimental correlation matrix shown in Fig. 4d. The relevant correlation coefficients can be extracted from the experimental correlation matrix shown in Fig. 4d, resulting in values of  $c_1 = 0.859$ ,  $c_{11} = 0.746$ ,  $c_2 = 0.532$ . ( $c_{11}$  was estimated by applying second-order polynomial interpolation to  $c_0 = 1$ ,  $c_1$  and  $c_2$ ). For these values, the noise transfer factor drops from  $\sqrt{20} = 4.47$  to  $\sqrt{0.613} = 0.783$ . As a result, the SD of the charge density calculated using Eq. (10) drops from  $0.64 \text{ e/p}^2$  to  $0.11 \text{ e/p}^2$ , in agreement with the experimental value of the charge density noise being  $0.10 \text{ e/p}^2$ .

Integration of the measured charge density distribution reduces noise, although it does not bring it back to the value that it had in the original phase image, both because the integration region is usually smaller than the FOV and as a result of noise correlations. (Even if the phase noise were uncorrelated, the charge density noise becomes correlated as a result of the use of the discrete Laplacian operator.) The right half of Fig. 4 illustrates how the SD of the measured charge  $\delta Q$  is related to  $\delta\sigma$  and  $\delta\varphi$ . We consider a simple square  $7 \times 7$  matrix with noisy pixels  $\delta\varphi$  representing the region of the phase image where we attempt charge measurement. The discrete Laplacian of this matrix, which is represented by a standard  $3 \times 3$  kernel, is an edge-padded  $5 \times 5$  matrix (the evaluation of boundary pixels is neglected for simplicity), in which each pixel value  $L_{ij}$  is a linear combination of the original pixel values written above (e.g.,  $L_{44} = \delta\varphi_{34} + \delta\varphi_{35} + \delta\varphi_{43} + \delta\varphi_{45} - 4\delta\varphi_{44}$ , etc.). Summing these 25 pixels gives another linear combination of pixel values. Isolating the coefficients of each pixel contributing to the sum and assembling them into a matrix yields what is shown in Fig. 4 at the end of the process diagram, which coincides with the discrete representation of the loop integral of the gradient (represented by the  $(-1, 1)$  kernel) of the original image over the boundary of the chosen region. In addition to providing visual verification of the equivalence of the two methods for measuring  $Q$ , it implies that  $\delta Q$  is also identical. Counting the number of pixels that contribute to the measurement of  $Q$  and summing them in quadrature (for uncorrelated phase noise) provides the following relationship between the measurement uncertainty and the phase noise SD:

$$\delta Q = \sqrt{8n} \frac{\epsilon_0}{C_E} \delta\varphi = \sqrt{\frac{2L}{p}} \delta q, \quad (11)$$

where  $n \gg 1$  is the number of pixels on one side of the square integration loop and  $L$  is the total length of the loop. The noise transfer factor in the case of correlated phase noise becomes  $\sqrt{8(1 - c_1)n}$ , where the only relevant correlations are those between adjacent pixels in the Laplacian. Since these correlations are also the strongest,  $c_1$  can

the specimen to have accumulated at the edges in the presence of the applied voltage of 50 V. The cumulative charge profile in the specimen (blue) is obtained by integrating the charge density in Fig. 7a parallel to the  $y$ -axis using the approach described in Section 2.1. The integrated charge is plotted by a red dashed rectangle in Fig. 7a. The charge in the specimen is negative. The approximately constant slope of the cumulative charge profile suggests that the charge density in each slice of the specimen is the same, *i.e.*, that it has an approximately constant linear charge density parallel to its axis.

#### Numerical model-dependent reconstruction

An analytical model-dependent approach for determining the charge density from an electron optical phase image relies on having access to a model that can be used to solve the Laplacian equation. A needle-shaped specimen has often been modelled as a line charge in the presence of a grounded conducting plane. The justification for using such a model is that equipotential surfaces around a line of constant charge density take the form of ellipsoids, which are in turn similar to the outer boundary of a needle-shaped specimen, which is often conducting and expected to be an equipotential. The charge density in such a model can be adjusted until a best match is found between experimental and simulated phase images in vacuum outside the specimen. The influence of the grounded conducting plane on the electrostatic potential distribution and its electron optical phase can be included by using image charge methods [22]. In the presence of an external field, a linear charge density that increases along the length of the needle can be used in the model instead of a constant charge density [23,46]. An analytical model for the electron optical phase [23] then takes the form

$$\varphi(x, y) = \frac{KC_E}{4\pi\epsilon_0} \left[ 4Ly + 4xy \arctan \frac{y-L}{x} - 4xy \arctan \frac{y+L}{x} - (L^2 + x^2 - y^2) \ln \frac{x^2 + (y-L)^2}{x^2 + (y+L)^2} \right] \quad (12)$$

In the present example, the shape of the needle was fitted to an ellipsoid of major semi-axis  $a = 45 \mu\text{m}$  and minor semi-axis  $b = 0.6 \mu\text{m}$ . The value of  $K$  in Eq. (12) was found from a best fit to the phase image. Fig. 8a shows a simulated contoured phase image corresponding to the best-fitting result, with  $K = 35 \text{ e}/\mu\text{m}^2$ . The electric field strength  $E$  induced by the counter-electrode and the base on which the field emitter sits can be calculated [23,46] and is approximately 0.4 MV/m, which is comparable to the electric field generated when 50 V is applied between two plates with a separation of 50  $\mu\text{m}$ . Fig. 8b shows a stream plot of the electric field lines in the  $z = 0$  plane for the best-fitting parameters. The maximum electric field strength at the apex  $E_{\text{apex}}$  is approximately 3.6 GV/m. The ratio  $E_{\text{apex}}/E$  corresponds to a field enhancement factor of 9000.

This field emitter was also transferred to a dedicated ultra-high-vacuum chamber to measure its field emission properties [48]. The electric field at the apex determined from a measured  $I$ - $V$  curve was 2.2 GV/m, which is slightly lower than that determined here. However, we did not detect a field emission current during our experiment, most likely due to the poorer vacuum level in the TEM column.

#### 4. Numerical iterative model-based reconstruction

The accuracy of the analytical model-dependent approach described

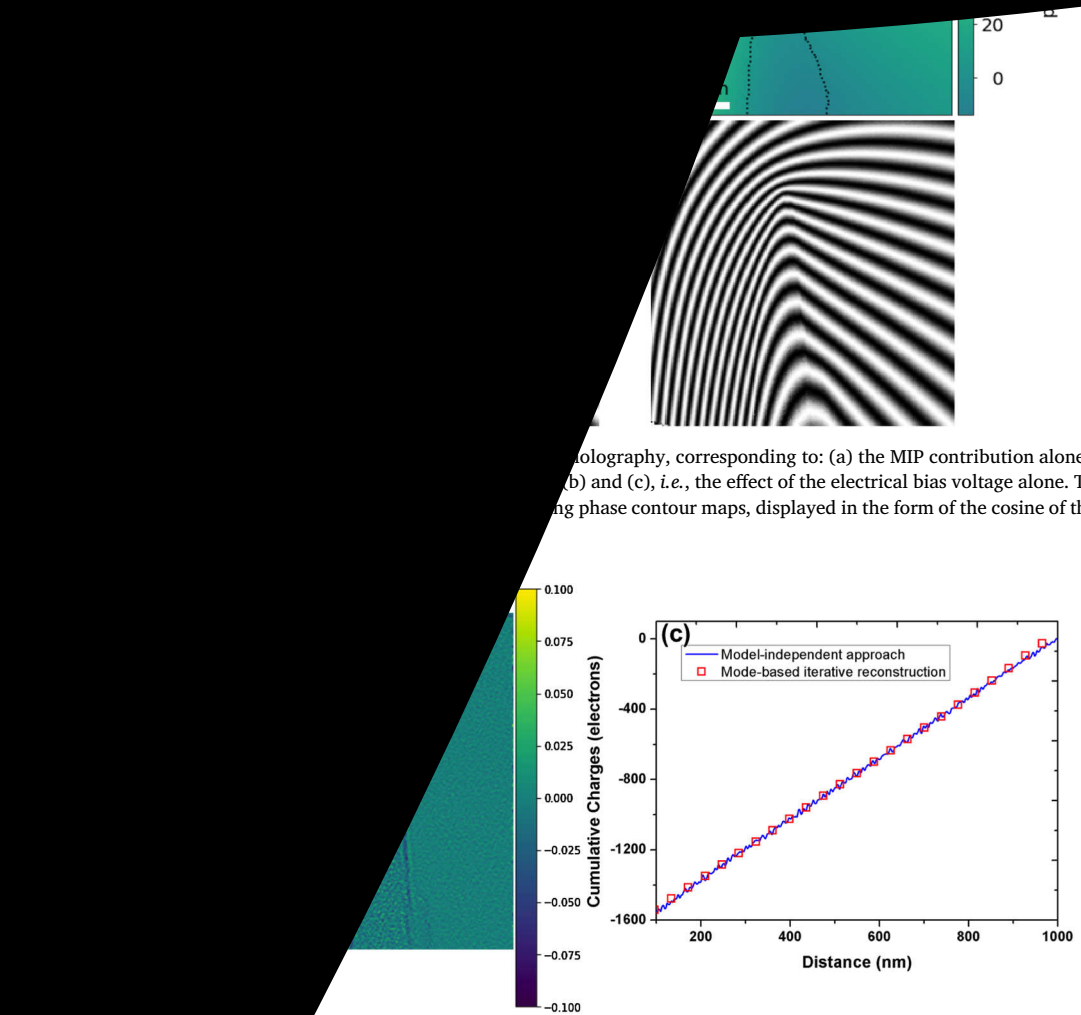
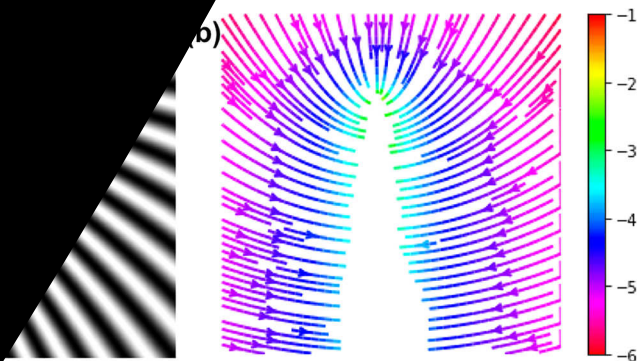


Figure 7. Comparison of experimental phase images and charge density profiles, corresponding to: (a) the MIP contribution alone; (b) the MIP contribution and (c), *i.e.*, the effect of the electrical bias voltage alone. The outline of the specimen is shown in the inset of (a). The phase contour maps, displayed in the form of the cosine of the phase. The charge density distribution is shown in the inset of (c).



**Fig. 8.** Demonstration of the application of the analytical model-dependent approach described in the text to a  $\text{LaB}_6$  field emitter that was electrically biased at 50 V. (a) Contoured phase image that provided a best fit in the vacuum region around the specimen to the experimental phase images shown in Fig. 6b and c. The phase contour spacing is  $2\pi$  radians. (b) Streamlines showing a section through the electric field lines in the  $z = 0$  plane calculated from the best-fitting parameters. The colour scale indicates the natural logarithm of the electric field strength. The shadow of the needle is indicated by the white region.

$$\left. \frac{1}{\sqrt{(x-x_0')^2 + (y-y_0')^2 + z^2}} \right\}. \quad (16)$$

the electrostatic potential in the  $z$  direction from  $+\infty$  to  $-\infty$  then the expression

$$\ln \frac{(x-x_0')^2 + (y-y_0')^2}{(x-x_0)^2 + (y-y_0)^2}. \quad (17)$$

It should be aware that two singularities are present in the above expression. To overcome this problem, we tackle this problem by treating a single voxel as a charged sphere. (See Appendix.) Fig. 10 shows a simulated phase image of a dipole, in which the two charges are separated by a distance of 32 nm.

Making use of Eq. (17), the forward model can be divided into a projection of a charge density distribution in the electron beam direction and a subsequent phase mapping operation described by a convolution. In discretised form, the projection and convolution can be executed in two steps, which can be described in matrix form by the expression

$$\mathbf{y} = \mathbf{F}\mathbf{x} = \mathbf{M}\mathbf{P}\mathbf{x}, \quad (18)$$

where the matrix  $\mathbf{F}$  is split into a projection matrix  $\mathbf{P}$  and a convolution matrix  $\mathbf{M}$ ,  $\mathbf{x}$  is the charge state vector (i.e.  $q(x, y, z)$  in vectorised form) and  $\mathbf{y}$  describes the calculated phase images in vectorised form. An efficient implementation of the projection step can be achieved by employing sparse matrix calculations, especially in the case of projections along  $z$  axis. In order to optimise the second step, the convolution kernels (see Eq. (17)) can be pre-calculated in real space and fast convolutions can be used in Fourier space [49,51].

#### 4.3. Regularisator

Regularisation provides a way of making use of *a priori* information in the model-based inverse algorithm. The following regularisators were used here: the application of a mask to indicate the region that can contain charges (i.e., the location of the specimen); the application of a confidence mask to define trustworthy regions in the experimental phase; and the enforcement of physical or mathematical constraints by adding a Tikhonov regularisator [52] in the cost function.

The total electrostatic potential energy  $W$  of all of the charges can be written in the form [53]

$$W = \frac{1}{8\pi\epsilon_0} \sum_i \sum_j \frac{q_i q_j}{|\mathbf{r}_i - \mathbf{r}_j|}, \quad (19)$$

where  $q_i$ ,  $q_j$  and  $\mathbf{r}_i$ ,  $\mathbf{r}_j$  are the magnitudes and positions of the  $i$ ,  $j$ th charges, respectively. The charges interact with each other through the Coulomb force, turning a linear term (Eq. (15)) into a non-linear one. For an ideal metal, in which charges are located only on the specimen surface, Eq. (19) can be reduced to the form

$$w = \frac{\epsilon_0}{2} \sigma^2, \quad (20)$$

where  $w$  is the energy density and  $\sigma$  is the surface charge density. Minimisation of the total potential energy is a physical constraint that can be used to restrict the reconstruction of the charges. In order to enforce this constraint, we use Tikhonov regularisation of 0th order, which corresponds to the use of a scaled identity matrix in the regularisation term. The regularisation term is then exactly Eq. (15), i.e., a Euclidean norm. Although, in general, charges do not need to be

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The reconstruction process is illustrated in Fig. 9. The workflow, which is used to map a chosen phase onto a corresponding electrostatic phase, is the basis for iterative reconstruction of the charge density. A simple and easy model that can be incorporated in the

quantities of charge density  $\mathbf{x}$  and phase  $\mathbf{y}$  should not be constrained in the  $(x, y)$  coordinates.

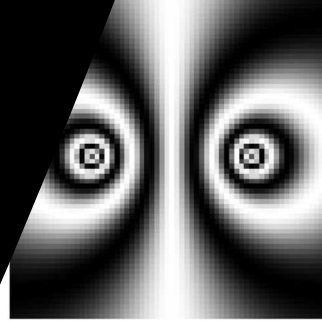


Figure 12: Calculated for a positive charge and its image charge separated by a distance of 150  $\mu\text{m}$ .

#### 4.4.1. Mask

A mask can be used to specify *a priori* information about the specimen geometry, *i.e.*, the positions at which charges can be located. It can therefore lead to a significant reduction in the number of unknowns and to an improvement in the quality of the reconstruction. Although masking could be implemented as a term in the cost function, here we apply a mask by excluding these regions from the charge state vector  $\mathbf{x}$ , which corresponds to assuming a charge of zero in these regions. The algorithm then does not fit the regions outside the mask. Differences between input and reconstructed projected charge density distributions obtained from the phase image shown in Fig. 12, both with and without using a mask, are shown in Fig. 13. When a mask is not used, the error in the reconstructed charge is approximately 10%, while that in the phase is more than 150  $\mu\text{rad}$  (not shown). There are also ripple-like artefacts in the reconstructed charge density. In contrast, when a mask (marked by the dashed circle in Fig. 12) is used, the error in the reconstructed charge falls to below 1%, while that in the phase falls to approximately 15  $\mu\text{rad}$  (not shown). In addition, the ripple-like artefacts are absent.

#### 4.4.2. Gaussian noise and regularisation strength

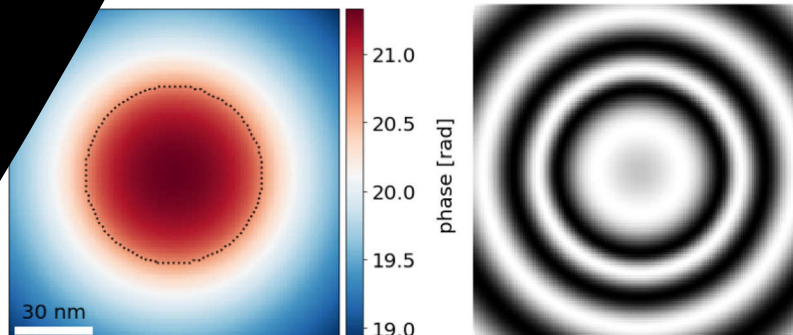
The noise in an experimental phase image can depend on the camera used and on the acquisition method (*e.g.*, single vs multiple hologram acquisition). In the presence of noise, reconstruction without using a regularisator is found to result in a charge density distribution that can deviate greatly from the input. A 0th order Tikhonov regularisator was therefore used here. As discussed in Section 4.1, the regularisation strength, which is defined by the value of  $\lambda$ , determines the ratio between the residual norm vector (the first term on the right of Eq. (14)) and the regularisation term. If  $\lambda \rightarrow 0$ , then the regularisation term vanishes and the cost function only relies on the residual norm vector, resulting in high frequency noise in the reconstructed charge density distribution. In contrast, if  $\lambda \rightarrow \infty$  then the cost function favours the regularisation term and the reconstructed result diverges from the experimental data. A good choice for the regularisation parameter corresponds to an optimal balance between compliance with the



reconstructed charge density distribution shown in Fig. 15. At the edge of the mask, the reconstructed charge density deviates by 50% from the input distribution, while elsewhere the error is below 5%. The corresponding phase error is below 1%. Fig. 16 shows the influence on the reconstructed charge density distribution of using regularisation parameters of 0.5 and 50. For  $\lambda = 50$ , the charge distribution is too smooth, whereas for  $\lambda = 0.5$  it is too noisy. It should be noted that the algorithm is designed to be insensitive to the presence of an arbitrary phase offset and an arbitrary phase ramp. Care in the interpretation of the result is therefore required if a real phase ramp may be present across the FOV.

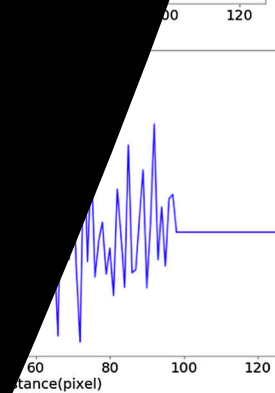
#### 4.4.3. Confidence mask

Experimental phase images can contain artefacts that originate from the specimen (e.g., unwanted effects of dynamical diffraction, contamination or electron-beam-induced charging), from the microscope (e.g., image distortions, or instabilities), from the detector (e.g., undersampling or dead pixels), or from image analysis (e.g., imperfect



phase image (left) and corresponding 8-times-amplified contour map (right) for the charge density distribution shown in Fig. 11. The phase is  $\frac{2\pi}{8}$  radians.

profiles across the centre ( $x = 0$ ) are shown on the right. In each case, 1000 iterations of the reconstruction algorithm were used.



density distribution is then determined only from the phase outside the charged sphere and can be seen to deviate significantly from the input charge density distribution. Interestingly, although the charge density distribution cannot be reconstructed reliably, the retrieved phase outside the charged sphere is consistent with the input phase, suggesting that the projected electric field can also be retrieved correctly outside the sphere. This is not surprising, since the algorithm always delivers a unique solution (in a mathematical sense) for a given type and strength of regularisation. However, without information about the phase in the interior of the object, reconstruction of the charge density inside it cannot be unique. For example, for a metallic ellipsoid the charge on the surface produces the same electric field distribution outside the object as a line of constant charge density located on its axis [22]. This statement is also consistent with the general property of a homogeneous Laplace equation that the values in the domain volume depend solely on the values or their derivatives on the domain boundary. The possibility of being able to reconstruct the potential and electric field outside a specimen without needing to make use of phase information inside it has significant implications for applications such as the characterisation of electrically biased needle-shaped specimens for atom probe tomography, for which the electric field outside the specimen rather than the charge density inside it may be the parameter of primary interest for providing experimental input for the simulation of ion trajectories. Nevertheless, whereas techniques based on differential phase contrast can be used to record the projected electric field directly, an argument in favour of reconstructing the charge density from an off-axis electron hologram before then using it to infer the projected electric field is that the charge density is localised within the specimen, rather than extending outside the FOV.

#### 4.4.4. Charges outside the field of view

As a result of the limited lateral extent of a phase image, it is often not possible to include all of the specimen or all of the phase change associated with the charge density distribution inside the FOV. This problem is particularly apparent when examining electrically-biased needle-shaped specimens, such as atom probe tomography needles or field emitters. It is illustrated for a uniform shell-like charge density distribution in Fig. 18. The phase of the entire shell (not shown) is calculated using the charge density shown in Fig. 18a. However, only

magnification, position or for example, it may not be possible to measure the phase everywhere across the specimen. In such cases, changes in electron-beam-induced phase contrast. For these reasons, a confidence mask is used to weight the signal in each pixel in the phase image. Pixels where the signal is trustworthy are assigned a value of 1, and pixels where it is not are assigned smaller values.

The weighted norm for the cost calculation of the reconstruction is defined by the matrix  $S_e^{-1}$  is diagonal, with each entry on the diagonal corresponding to a single pixel in each phase image. These entries are the values of the confidence matrix. If the confidence value is zero, then the corresponding residual does not contribute to the cost function. The weighted norm therefore takes the

result of a reconstruction performed after assigning a weight of 1 to the pixels in the charged region (i.e., inside the specimen) and 0 to the pixels outside. The reconstructed projected charge

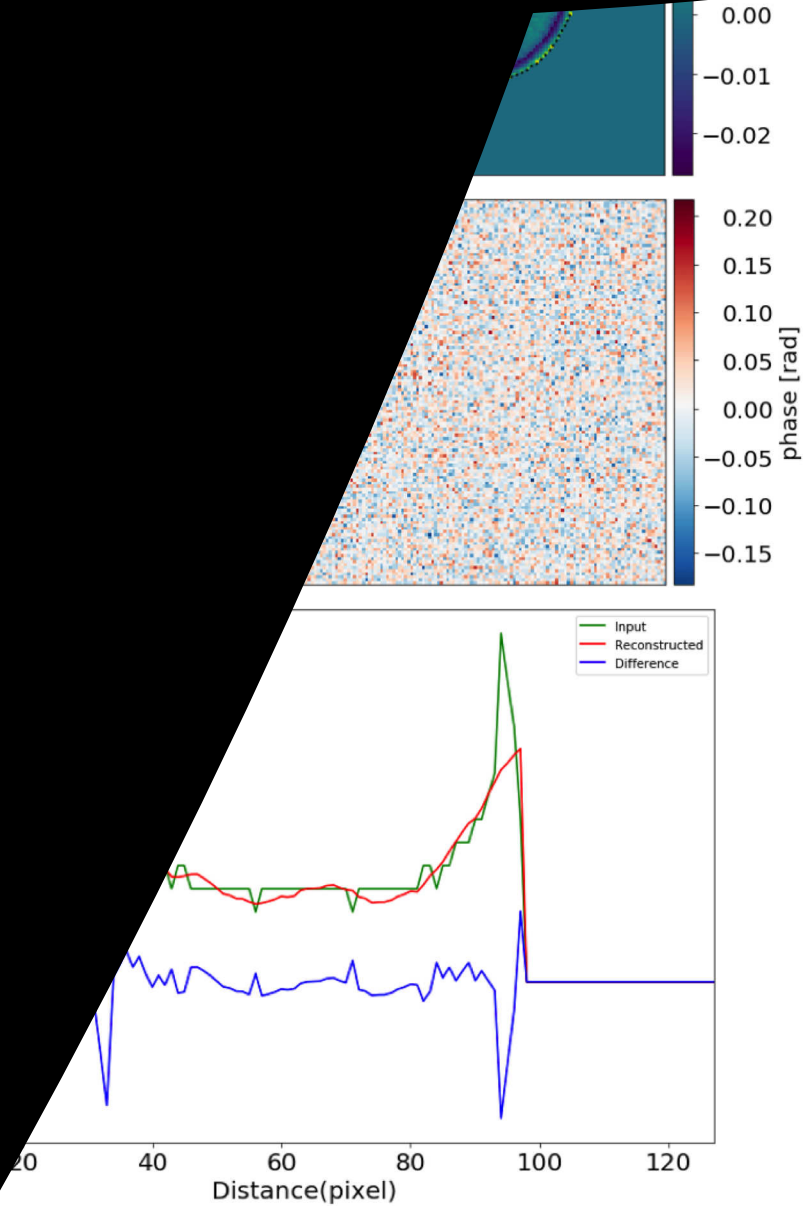


Figure 12: The phase image shown in Fig. 12 with Gaussian noise of 0.05 rad added, for a regularisation parameter  $\lambda$  of 5, showing the charge density distribution and its deviation from the input charge density (upper row, shown in units of e/pixel), the reconstructed phase (middle row), and the charge profile across the centre of the image ( $x = 0$ ) extracted from the reconstructed (red) and input (green) charge densities (lower row). The blue line shows the difference between the reconstructed and input charge densities.

The input charge density was used for the reconstruction. The reconstructed charge density distribution, which is shown in Fig. 12b, is significantly different from the input charge density, as the input charge density is present only in the masked region, while the reconstructed charge density also contributes to the phase. Since the input charge density does not include any boundary conditions, with the reconstructed charge density, the presence of charges outside the FOV is taken into account by making use of additional buffer pixels, which are located outside the edge of the FOV. These buffer pixels can be used to

introduce a distribution of additional charge density around the edge of the image, in order to take into account the influence of unknown charges outside the FOV. They are only used during the reconstruction and are discarded when displaying the final reconstructed charge density inside the part of the specimen that is within the FOV. In the present example, Fig. 18c shows that the reconstructed result is almost consistent with the original input charge density when 8 buffer pixels are used at the border of the image.

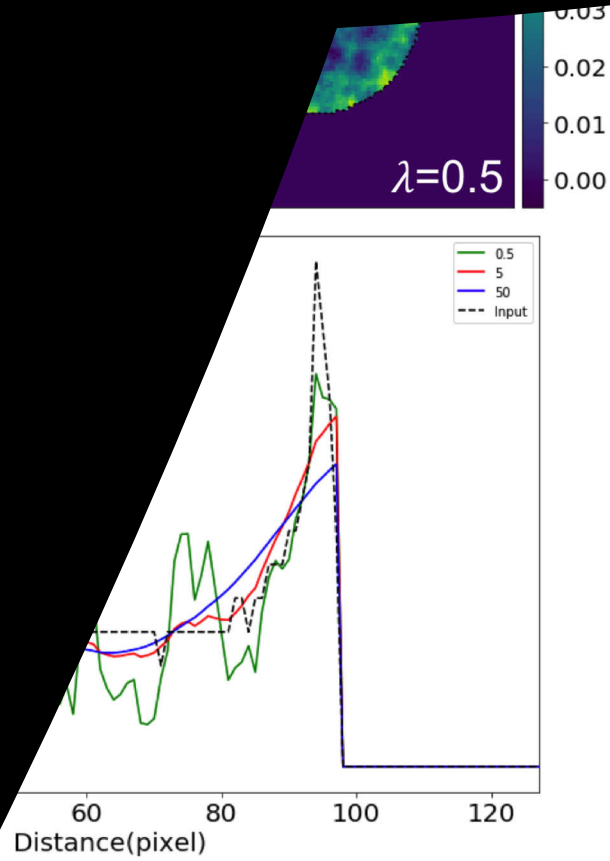


Figure 19: Charge density distributions starting from the phase image shown in Fig. 12 for regularisation parameters  $\lambda$  of 0.5, 5 and 50. The top panel shows the charge profiles across the centre for the input charge distribution (black) and the reconstructed charge density for regularisation parameters: 0.5 (green), 5 (red), and 50 (blue).

The PRW can affect the reference wave that is disturbed significantly by the electric fields, which may influence the PRW. The influence of the PRW can be taken into account by implementing a modified kernel. A schematic illustration of the modified kernel is shown in Fig. 19 for a point charge within the FOV. In the region of the tail of the electric field arising from the point charge, the measured phase image is asymmetrical with respect to its position. The projected potential of the point charge in the vacuum reference wave originates from the electric field of the point charge, which decays with distance. The solid red line in Fig. 19 represents the projected potential of the point charge in the vacuum reference wave region, which has to be added to the input phase image with a negative sign to take the PRW into account. The resulting phase image is asymmetrical with respect to the

position of the charge. The red dashed line can be described as originating from a negative (virtual) point charge located on the other side of the biprism, as shown by a solid blue line in Fig. 19. The influence of the PRW can therefore be described by a region of virtual charges of opposite sign (mirror charges) that are located on the opposite side of the biprism at a distance that is equal to the interference distance. It can be treated in the same way as any other source of charge located outside the FOV, as described in Section 4.4.4, or alternatively by using a modified kernel that includes the PRW effect.

#### 4.4.6. Reconstruction from an experimental phase image

A phase image recorded from a LaB<sub>6</sub> needle-shaped specimen that was electrically biased *in situ* in the TEM using an applied voltage of 50 V (Fig. 6c), from which the MIP contribution to the phase had been removed using the procedure described above, was used for reconstruction of the charge density. The positions of image point charges in the forward model were calculated by assuming a distance between the needle and the contour-electrode of approximately 6  $\mu\text{m}$ . The sampling density was 7 nm/pixel. A 4-pixel-wide buffer was also defined around the border of the image, in order to compensate for the presence of charges at unknown positions outside the FOV, as well as the presence of the PRW, as described in Sections 4.4.4 and 4.4.5. 0th order Tikhonov regularisation was used. An optimal value for the

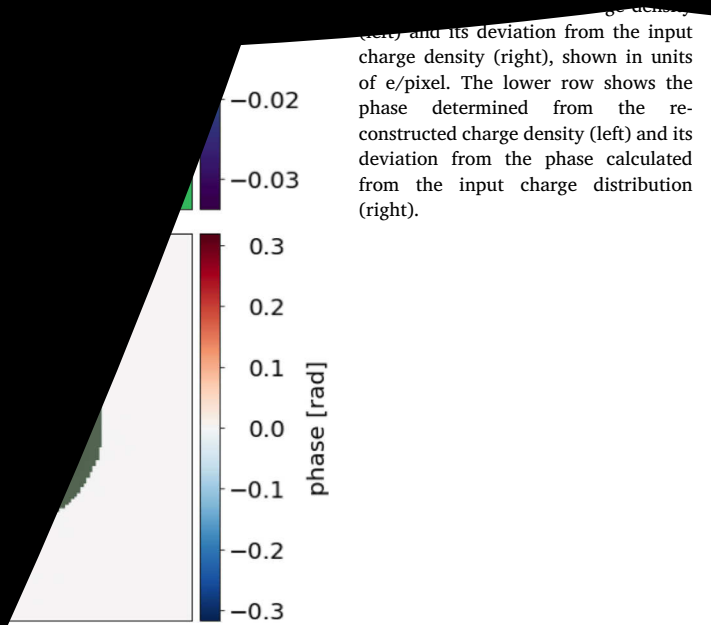


Figure 18. Illustration of the reconstruction of only part of a charge density distribution to highlight the influence of the presence of charges outside the FOV. (a) shows a uniform shell-like charge distribution, which was used to generate a phase image. (b) and (c) show reconstructed charge distributions generated from only the left half of the phase image. (b) was generated without using any boundary pixels. (c) was generated by allowing additional charge density to be fitted in an additional boundary region that had a width of 8 pixels just outside the right edge of the image. The boundary pixels were then removed to display the final fitted charge density within the original FOV. All of the images are shown in units of  $e/\text{pixel}$ .

original experimental image (see Fig. 20c and d), this discrepancy is at a level of below 1% and may result from a slight error in the selection of the mask, or from the finite sampling of the phase at the narrow apex of the needle. If the reconstructed charge density distribution obtained using the MBIR algorithm (Fig. 20) is compared with that obtained from the Laplacian of the phase (Fig. 7), it is clear that the noise in the fitted charge distribution is greatly reduced, as a result of the use of *a priori* knowledge (in particular, the mask and the regularisation parameter) when performing the reconstruction. Nevertheless, it should be noted that the result depends strongly on the values of several input parameters, which should be chosen carefully when applying the MBIR approach.

#### 4.5. Reconstruction of charge density in three dimensions

Three-dimensional charge density distributions can in principle be

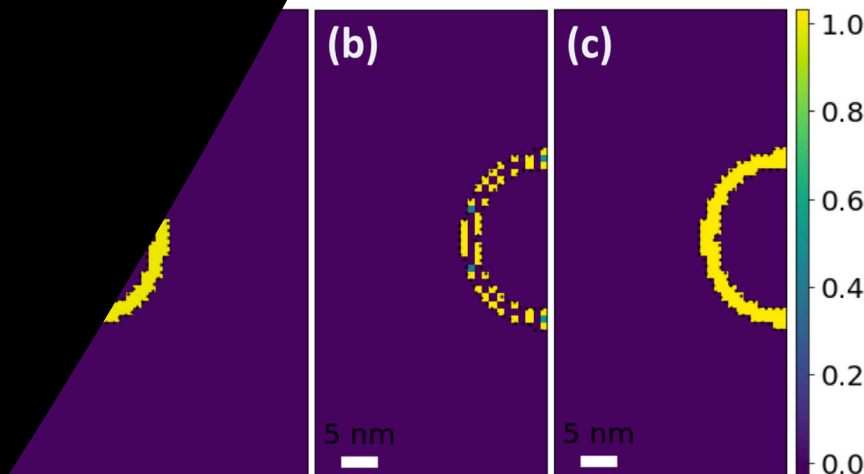


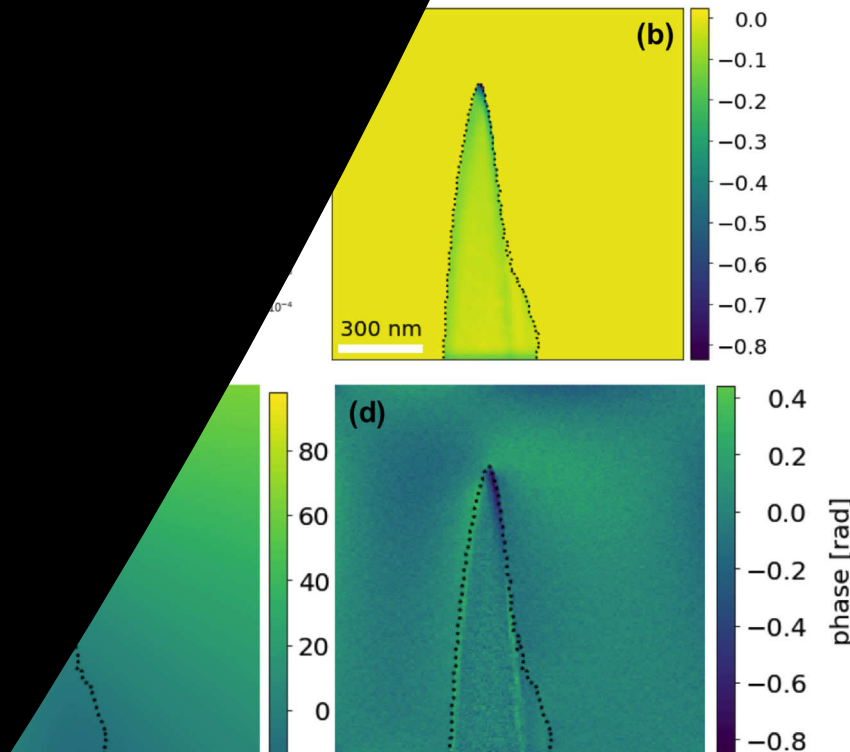
Fig. 18. Illustration of the reconstruction of only part of a charge density distribution to highlight the influence of the presence of charges outside the FOV. (a) shows a uniform shell-like charge distribution, which was used to generate a phase image. (b) and (c) show reconstructed charge distributions generated from only the left half of the phase image. (b) was generated without using any boundary pixels. (c) was generated by allowing additional charge density to be fitted in an additional boundary region that had a width of 8 pixels just outside the right edge of the image. The boundary pixels were then removed to display the final fitted charge density within the original FOV. All of the images are shown in units of  $e/\text{pixel}$ .

...phase image of a needle-shaped  $\text{LaB}_6$  specimen that was electrically biased *in situ* in the TEM at 50 V, from which the MIP contribution to the phase had been subtracted (Fig. 6c), using 0th order Tikhonov regularisation. The tilt angle range was chosen to be  $\pm 50^\circ$ . The tilt angle range can be achieved experimentally. The angle range was chosen to be  $10^\circ$ , resulting in an input dataset to the algorithm of 11 phase images in total. Three different three-dimensional masks were used: the shell, the outer surface of the sphere and the three-dimensional volume. 0th order Tikhonov regularisation was used, as described in Section 4.3. The regularisation parameter  $\lambda$  was set to be 100 for all three cases.

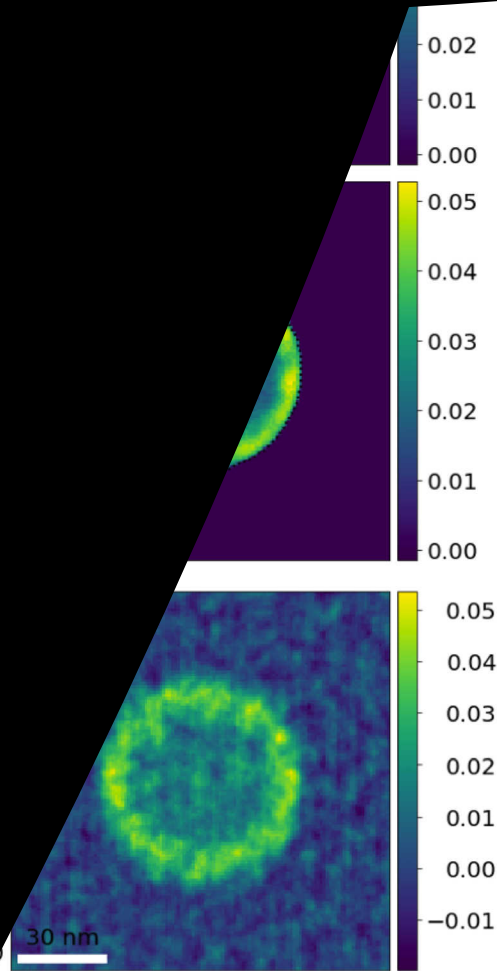
Figs. 20 and 22 show two-dimensional slices and line profiles of the reconstructed three-dimensional charge density. As expected, the use of a shell mask, which defines the true positions of the charge, delivers the best results. If only the outer surface of the sphere is used as a mask, then the algorithm retrieves the key feature of the charge distribution (the homogeneous surface charge) correctly. However, the reconstructed charge is smoothed slightly into the volume of the sphere and exhibits small oscillations next to the shell region. If the three-dimensional volume is used, then the basic features of the charge density are reproduced (see Fig. 22), but additional spreading of the charge and high frequency artefacts are present across the FOV. Although further tests are required to optimise the use of the MBIR approach for three-dimensional charge density reconstruction, the results presented here are highly encouraging.

## 5. Summary and conclusions

Three different approaches have been described for the measurement of charge density distributions in nanoscale materials from electron optical phase images recorded using off-axis electron holography:



**Fig. 20.** Reconstruction of the projected charge density distribution from an experimental phase image of a needle-shaped  $\text{LaB}_6$  specimen that was electrically biased *in situ* in the TEM at 50 V, from which the MIP contribution to the phase had been subtracted (Fig. 6c), using 0th order Tikhonov regularisation. (a) Application of L-curve analysis to determine that the optimal value of the regularisation parameter  $\lambda$  is 10. (b) Reconstructed projected charge density distribution, shown in units of e/pixel. (c) Phase image determined from the reconstructed projected charge density. (d) Difference between the reconstructed phase shown in (c) and the experimental phase image shown in Fig. 6c. Note the different intensity scales in (c) and (d).



parameter  $\lambda$  is set to be 100 for all three cases. Different three-dimensional masks were used to define the possible locations of the reconstructed charge: the shell that defines the original charge distribution (upper row); the outer surface of the sphere (middle row); the full three-dimensional reconstruction volume (lower row). The left column shows the reconstructed charge distribution in the central slice ( $z = 0$ ), while the right column shows the corresponding projected charge density distribution.

approach (Section 3.2), in which a model is used to simulate phase images from a recorded phase image; the model-independent approach (Section 4), in which a model is used to simulate phase images from a recorded phase image; the model-dependent approach (Section 3.2) relies on the specimen geometry and requires the inclusion of a reference wave to be included in the model. The model-independent approach (Section 3.1) is the most direct and straightforward, as it only requires the projected charge density distribution to be used and is insensitive to the presence of a reference wave and charges outside the field of view. However, the charge density can be noisy (since the approach involves the use of derivatives). In the MBIR approach (Section 4.4.6), each charged voxel is approximated as a homogeneous sphere and a mirror charge. It can incorporate a reference wave through the use of masks, regularisation parameters and other physical constraints, resulting in lower noise but requiring

care in the selection of parameters to avoid the introduction of artefacts. A further advantage is that boundary pixel regions can be used to take account of the presence of charges outside the field of view and the perturbed reference wave (Section 4.4.4). Artefacts can be tackled by assigning zero confidence to regions of phase images that contain untrustworthy information (Section 4.4.3). It is important to note that different charge distributions inside an object can result in the same electrostatic potential and phase distribution outside it.

The three approaches have been tested on an experimental phase image of an electrically biased needle-shaped LaB<sub>6</sub> specimen and have been shown to provide consistent results for the charge density. The phase shift of a line charge is used as a simple model in the analytical model-dependent approach. Projected charge density distributions retrieved using the model-independent approach (Fig. 7a and b) and the MBIR approach (Fig. 4.4.6) show that most of the charge is located close to the surface of the needle, with charge accumulation at its apex. The result obtained using the MBIR approach has much less noise than that obtained using the model-independent approach. Three-dimensional charge density distributions can in principle be reconstructed using each approach, either by applying a standard backprojection-based tomographic reconstruction algorithm to projected charge

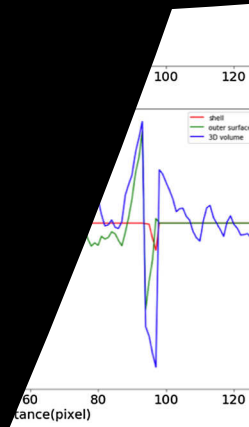


FIG. 3. Line profiles of the electrostatic potential  $V(x, y, z=0)$ . The three line profiles correspond to reconstructions for: “3D volume” – no mask applied (blue), “outer surface” – mask includes the whole sphere (green), “shell” – mask includes only the shell of the sphere where the charge was placed (red).

with for focused ion beam preparation of the  $\text{LaB}_6$  specimen. This project was carried out within the framework of a scientific service agreement between the Ernst Ruska-Centre for Microscopy and Spectroscopy with Electrons in Forschungszentrum Jülich and Gatan Inc. The authors acknowledge the European Union for funding through the Marie Curie Initial Training Network SIMDALEE2 (Marie Curie Initial Training Network (ITN) Grant No. 606998 under FP7-PEOPLE-2013-ITN). V.M. thanks the Deutsche Forschungsgemeinschaft for funding within the framework of the SFB 917 project “Nanoswitches”. R.D.-B. thanks the Deutsche Forschungsgemeinschaft for a Deutsch-Israelische Projektkooperation (DIP) Grant and the European Union's Horizon 2020 Research and Innovation Programme Q-SORT (Grant No. 766970 under H2020-FETOPEN-2016-2017). This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No. 823717-ESTEEM3.

radius  $R$  and charge  $q$  can be written in the form

$$d_1 \geq R$$

$$d_1 < R$$

$$\text{if } d_2 < R$$

$d_2 = \sqrt{(x - x'_0)^2 + (y - y'_0)^2}$  are the projected distances to the charge and its image charge, respectively, the heights at which an electron enters the sphere and its image, respectively, and  $(x_0, y_0)$  and  $(x'_0, y'_0)$  are the coordinates of the charge and its image, respectively.

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