

Matlab Imaging Algorithms: Image Reconstruction, Restoration, and Alignment, with a Focus in Tomography. (Version 2.2)

Toby Sanders
email: toby.sanders@asu.edu
School of Math and Stat Sciences,
Arizona State University, Tempe, AZ.

Abstract

This document is a guide for several MATLAB imaging code, which can be used for image reconstruction, denoising, alignment, etc. There is a general focus on tomographic reconstruction and alignment, as well as ℓ_1 regularization algorithms.

1 Solvers for Signal, Image, and Volume Reconstruction

1.1 Higher order TV based ℓ_1 Regularization

- **Code name:** HOTV3D
- **Demos:** see demo_L1optimization_simple.m, demo_tomo.m, demo_MHOTV.m, demo_inpaint.m, and demo_1D_L1.m
- **References:** [6, 1, 9, 4, 7]
- **Description:** An iterative solver for higher order total variation ℓ_1 regularization minimization for inverse problems and denoising. MATLAB comments:

```
% Modifications by Toby Sanders @ASU
% School of Math & Stat Sciences
% 08/24/2016
%
%
% This code has been modified to solve l1 penalty problems with
% higher order TV operators. Several small bugs and notation
% changes have been made as well.
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%           Problem Description           %
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% function [U, out] = HOTV3D(A,b,n,opts)
%
% Motivation is to find:
%
%           min_f { mu/2*||Au - b||_2^2 + ||D^k u||_1 }
```

```

% where D^k is kth order finite difference.
% Multiscale finite differences D^k can also be used.
% To see how to modify these settings read the file "check_HOTV_opts.m"
%
% The problem is modified using variable splitting
% and this algorithm solves:
%
%      min_{u,w} {mu/2 ||Au - b||_2^2 + beta/2 ||D^k u - w ||_2^2
%                + ||w||_1 - (delta , Au - b ) - (sigma , D^k u - w) }
%
% delta and sigma are Lagrange multipliers
% Algorithm uses alternating direction minimization over f and w.
%
%
% This algorithm was originally authored by Chengbo Li at Rice University
% as a TV solver called TVAL3.
% original code and description can be found here:
% http://www.caam.rice.edu/~optimization/L1/TVAL3/
%
% Inputs:
%   A: matrix operator as either a matrix or function handle
%   b: data values in vector form
%   n: image/ signal dimensions in vector format
%   opts: structure containing input parameters,
%         see function check_HOTV_opts.m for these
%
%
% Outputs:
%   U: reconstructed signal
%   out: output numerics

```

- **Important fields in the opts structure:**

mu primary parameter balancing the data and regularization terms. This parameter is not as important if the data multiplier δ is implemented and updated a sufficient number of times (see `opts.data_mlp` term below).

beta secondary parameter to encourage $D^k f = w$. This parameter is not as important if a sufficient number of outer iterations are used (see `opts.outer_iter` below).

order order (k) of the finite difference operator used in the regularization term, can be set to any real number ≥ 0 . Generally recommended between 1 and 3.

levels Set to an integer generally between 1 and 4. For values greater than 1 a multiscale regularization approach is implemented.

data_mlp set to true or false, specifying whether or not the Lagrangian multiplier δ is used and updated. This parameter has a major effect. Set to true to approximately solve the constrained problem $Af = b$, or to simply accelerate convergence on the data term. Can also set to a positive integer value to indicate the maximum number of updates for δ .

L1type set to 'isotropic' to solve the isotropic model or 'anisotropic' to solve the anisotropic model. Isotropic is generally preferred.

outer_iter number of iterations or updates on the Lagrangian multiplier terms (as discussed above, the updates on δ are also limited by `opts.data_mlp`).

inner_iter number of iterations between each update on the Lagrangian multiplier terms, thus the maximum number of total iterations is `inner_iter*outer_iter`

tol_out outer loop tolerance used to terminate the algorithm

tol_inn inner loop tolerance used to move to next outer loop

nonneg if set to true, problem is solved under the constraint $f \geq 0$

max_c if set to true, problem is solved under a max constraint, $f \leq \text{max_v}$

max_v maximum value used if `max_c` is set to true

wrap_shrink ... set to to false unless signal is periodic.

** Other options are available, a full description of each is given in the “check_HOTV_opts.m” file. It is generally recommended to just use default values.*

-For automatic HOTV reconstruction of tomographic data see the code *HOTV3D_tomo.m*.

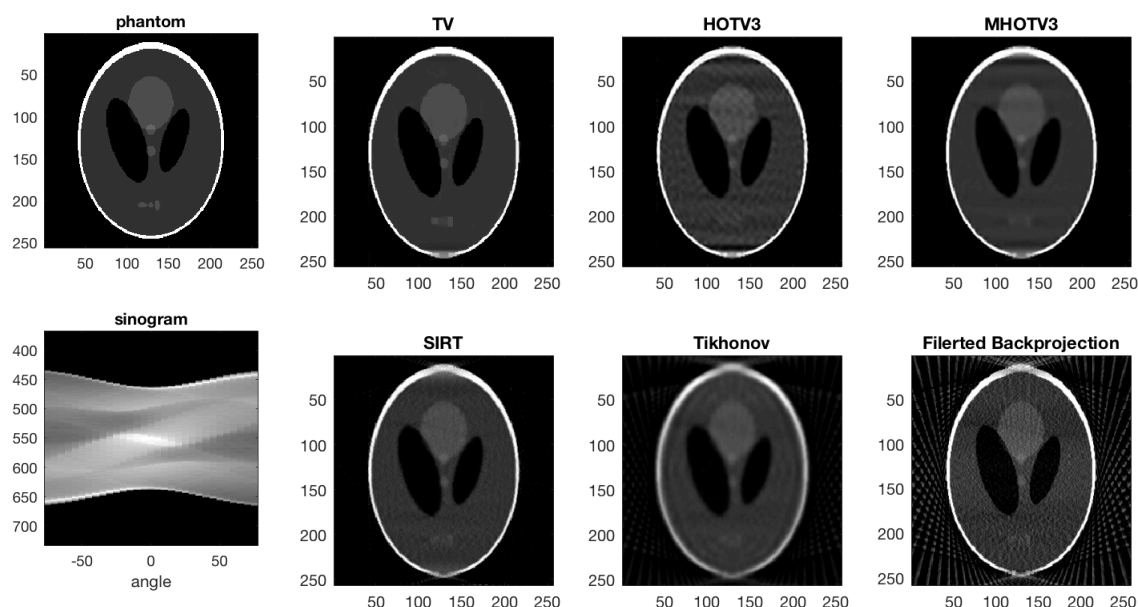


Figure 1: Tomographic reconstruction of phantom image.

1.2 General ℓ_1 regularization algorithm

- **Code name:** `lloptimo`
- **Demo:** see `demo_MHOTV.m`
- **Description:** An iterative solver for general ℓ_1 regularization minimization for inverse problems and de-noising.

```

% function [U, out] = l1optimo(A,b,n,opts)

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%           Problem Description           %
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% Motivation is to find:

%           min_f { mu/2*||Af - b||_2^2 + ||D f||_1 }

% This algorithm is essentially the same as HOTV3D, however,
% the user must specify the regularization operator "D". To do this,
% the user should input into the opts structure the options opts.D and
% opts.Dt, where opts.D is the regularization transform, and opts.Dt is
% the adjoint of opts.D.

```

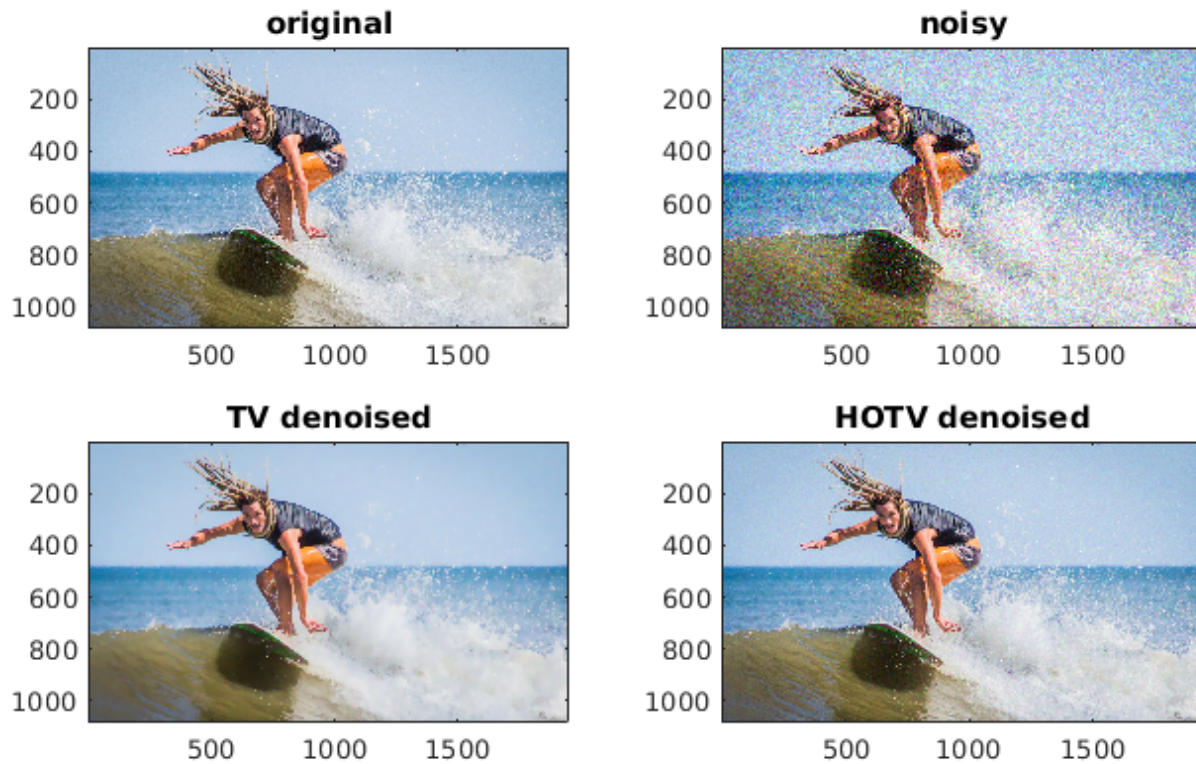


Figure 2: Denoising with ℓ_1 regularization.

1.3 Tomographic reconstruction with SIRT

- Code name: SIRT

- **Demo name:** see `demo_tomo.m`
- **References:** [11]
- **Description:** Simultaneous iterative reconstruction technique (SIRT) for tomographic reconstruction, which works by the general gradient decent method. MATLAB comments:

```
% function [x,out] = SIRT(stack,angles,recsize,iter,minc,maxc)
%
%DESCRIPTION:
%   %this function performs the SIRT algorithm for tomographic reconstruction.
%
%NOTATION:
%   % function [x,out] = SIRT(stack,angles,recsize,iterations,minc,maxc);
%
%INPUTS:
%   % stack - the tilt series, where it is assumed the tilt axis is horizontal
%   %   and located at the middle of the stack
%   % angles - a vector holding the projection angles of the stack, in order,
%   %   in degrees
%   % recsize - the dimension of the reconstruction
%   % iter - the number of SIRT iterations
%   % minc - minimum density constraint, e.g.  $U \geq 0$ .
%   % maxc - maximum density constraint, e.g.  $U \leq 1$ .
%   %
%   % default values are used for recsize, iter, and minc if they are not
%   % specified, therefore one may simply input "sirtden(stack,angles)."
%   % the recsize will be set to the detector count, i.e. size(stack,1)
%   % 50 iterations is default, and no density constraint is used if it is not
%   % specified
%
%OUTPUT:
%   % x - the reconstruction from the input tilt series and other input
%   %   parameters
%   % out - additional outputs
```

- **Additional Notes:** For automated reconstruction of a 3D volume see the algorithm *sirtauto.m*.

1.4 Discrete tomographic reconstruction with DART

- **Code name:** DIRT
- **Demo name:** `demo_DART.m`
- **References:** [2, 3, 8, 5]
- **Description:** Discrete algebraic reconstruction technique for tomographic reconstruction with prescribed gray levels, designed and extensively implemented by K.J. Batenburg et. al. MATLAB comments:

```

% function [U,out,init] = DIRT(opt,bb,init)

%INPUTS:
%Inputs that must be user defined:
% bb - input projection data, can be left empty if user instead
% specifies opt.data_name to reference saved data
% init - initial solution, can be left empty and algorithm will
% automatically compute a SIRT solution for initialization
% fields in the opt structure:
%opt.angles - the projection angles listed in order, in degrees
%opt.thresh - a vector holding the thresholds used for segmenting
%opt.grays - a vector holding the gray values used in segmentation.
%opt.data_name - name of the file in which the aligned tilt series is
% saved. Note the tilt axis should be horizontal and centered.
%opt.data_type - type of file of opt.filename, set to either 'mrc' or
% 'mat'
% optional fields for the opt structure that have default values if left
% unassigned
%opt.inner_iter - number of SIRT iterations in each DART iteration.
% Default value is 10.
%opt.W - if the projection matrix is precomputed, set opt.W to be
% the projection matrix.
%opt.disp - prints information about the reconstruction at each
% iteration. Set to "true" for the display and "false"
% otherwise. Default is "true".
%opt.disppic - displays various images about the reconstruction at
% each iteration. Set to "true" for this display and "false"
% otherwise. Default is "true".
%opt.convergence_criteria - if set to "true", DART will iterate until
% the reconstruction converges. Set to "false" to simply use
% opt.outeriter instead. Default is "false". Typically
% convergence is seen after just a few iterations.
%opt.convergence_tol - tolerance for opt.convergencecriteria.
% Default is 10-3.
%opt.max_iter - maximum number of iterations if
% opt.convergencecriteria is used. Default is 50.
%opt.rx - radius of the gaussian in the slice (x,y) dimension used
% for smoothing. Set to 1 for no smoothing. Default is 3.
%opt.rz - radius of the gaussian in the slice z dimension used
% for smoothing. Set to 1 for no smoothing. Default is 3.
%opt.sigmax - sigma value used for the gaussian smoothing in the
% (x,y) dimension. Default is 1. Increase for more smoothing
% and decrease for less smoothing.
%opt.sigmaz - sigma value used for the gaussian smoothing in the
% z dimension. Default is 1.5. Increase for more smoothing
% and decrease for less smoothing.
%opt.chunksize - size of the chunks that DART will run on. Default

```

```

    %is 50.
    %opt.overlap - once the chunks are computed, they are merged
    %together using a partition of unity. The overlap is the overlap
    %of this merging. Default is 10.

```

%OUTPUTS:

```

    %U - the 3-D DART reconstruction. The slices change with the 3rd
    %dimension, i.e. dartrec(:,:,i) holds the ith slice of the
    %reconstruction
    %out - data about the reconstruction
    %init - the SIRT reconstruction or input reconstruction used as
    %the initial solution to compute U.

```

%Inputs recommended to be user defined, but default values are otherwise
%used:

```

    %opt.bdry_dirt_iter - number of DART iterations with boundary
    %refinements
    %opt.region_dirt_iter - number of region refinements (DIP-LS)
    %opt.t_tol - radius of interval used for the partial segmentation
    %function with DIPS. Set a small tolerance for slower more
    %careful convergence and a high tolerance for fast convergence
    %opt.t_delta - increase in opt.t_tol whenever new pixels are not being
    %classified
    %opt.t_epsilon - this epsilon is used to determine if opt.t_tol should
    %be increased by opt.t_delta. It is the case if the number of
    %classified pixels has only changed relatively by less than
    %epsilon.
    %opt.bdry_update_type - indicates the solver used for bdry updates.
    %Set to 'SIRT' (default) or 'CGLS'.
    %opt.region_update_type - indicates the solver used for region updates.
    %Set to 'SIRT' (default) or 'CGLS'.
    %opt.rand_pixel_probability - probability to a free pixel is set to
    %free. Default is 0.
    %opt.resolution - the number of pixels used for reconstruction.
    %Reconstruction slices will be opt.recsz by opt.recsz. Default
    %value is the detector count
    %opt.startslice - which slice the reconstruction begins on. Default is 1.
    %opt.endslice - which slice the reconstruction ends on. Default is the
    %end of the stack.
    %opt.initialsolution - Set to "true" an initial solution has been
    %computed and will be used, otherwise set to "false".
    %opt.initialfile - name of the matlab file where the reconstruction is
    %saved

```

2 Image Alignment

2.1 Basic Cross Correlation

- **Code names:** cross_corr.m and cross_corr_pad.m
- **Demo name:** see demo_align.m
- **Description:** these algorithms perform 2D image alignment from an input 3D stack of images by cross correlation. “cross_corr.m” maintains the original image size by simply rotating or circular shifting the image and “cross_corr_pad.m” pads the stack with zeros based on the minimum and maximum shifts. MATLAB comments:

```
%DESCRIPTION
    %This function performs cross correlation of an input stack of images.
    %
%NOTATION
    %[stacknew,shifts]=cross_corr(stack,startslice);
    %
%INPUTS
    %stack - 3D matrix assumed to be a stack of images to be cross
        %correlated
    %startslice - the reference image that all slices are consecutively
    %aligned to. This image will not move
    %
%OUTPUTS
    %stack- 3D stack of images after cross correlation
    %shifts - shifts that were applied to each image
    %
%NOTE: This function automatically uses a window to improve alignment. To
    %remove the windowing, set pwr=0.
```

2.2 Center of mass alignment for tilt series alignment

- **Code names:** COM_align.m
- **Demo name:** see demo_align.m
- **References:** [10]
- **Description:** Center of mass alignment for electron tomography data. See <http://ascimaging.springeropen.com/articles/10.1186/s40679-015-0005-7> for details. MATLAB comments:

```
% Center of mass alignment for electron tomography data
%
% function [stack,shifts,usables] = COM_align(stack,angles,ratio,s)
%
% Inputs: stack is 3-D matrix containing tomography data
% angles are the angles in degrees
% ratio is the ratio of slices to be used for each projection
```



```

% s is the number of projections used as a sequence to determine rigid
% alignment. Setting s to be the number of angles is equivalent to the
% alignment method published here:
% http://ascimaging.springeropen.com/articles/10.1186/s40679-015-0005-7
%
%
%
% The ratio variable:
% should be input as a number between 0 and 1, where the user rates the
% consistency of the stacks through the slices. From our experience, very
% small values of s are better. If the user does not input a
% rating, the default value of .1 is used. Setting ratio
% to perfect 1 would mean that through the slices, there is significant amount
% of mass in all of the slices, and very little mass moves in or out as the
% tilt angle changes. A lower ratio would mean otherwise.
% This rating is then used to determine how many slices to use for the
% center of mass alignment. For example, if you set the rating to .2, then
% only the best 20 percent of the slices will be used in the alignment step.
% That way, if the the stack has some slices with very little mass, yet
% mass moves into these slices at high angles, these slices may be ignored
% for the alignment.

```

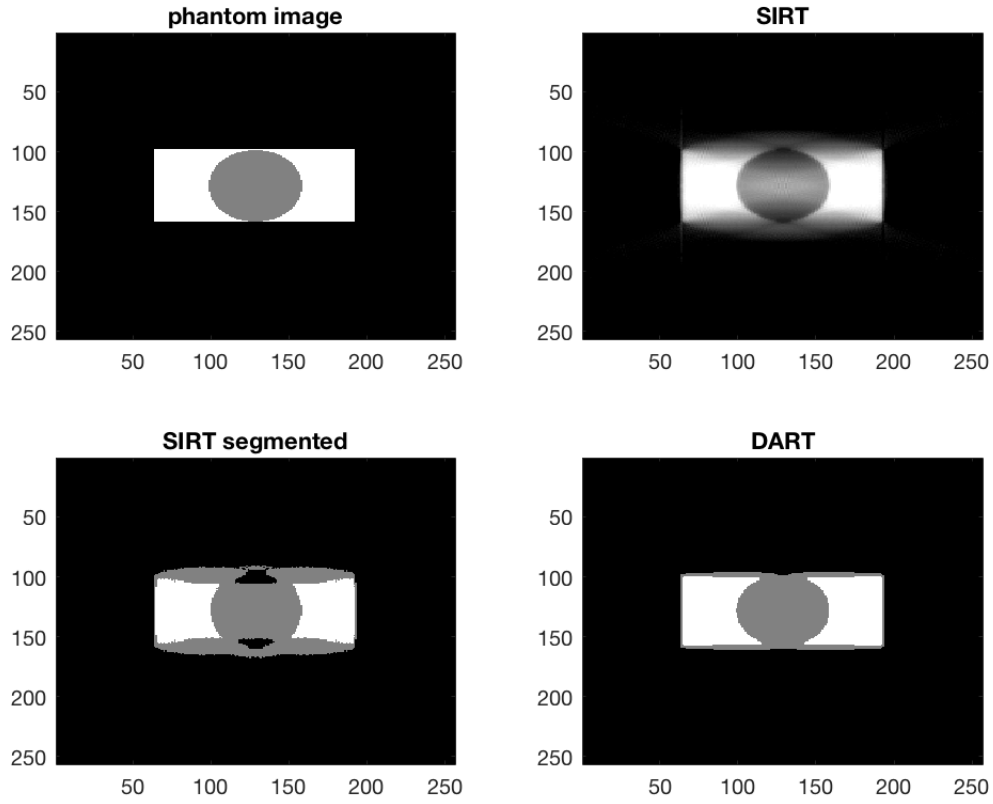


Figure 3: Tomographic reconstruction with DART.

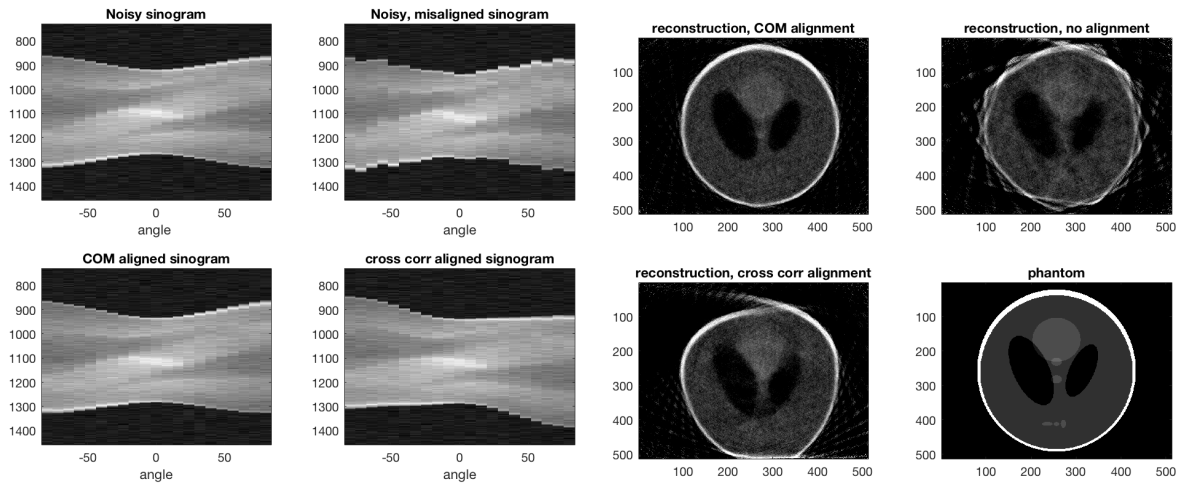


Figure 4: Tomographic alignment and reconstruction (SIRT).

3 Other utilities and codes potentially of interest

**See the comments within each code for details.*

- moviestack.m - used to visualize a 3D volume by passing through 2D cross-sections.
- radonmatrix.m - builds a sparse projection matrix for tomography based off of input geometry.
- corrx_global.m - a horizontal image alignment procedure for a tilt series based off of the conservation of mass.
- manualalign.m - a utility that can be used to manually align images.
- forward_proj.m - builds a tilt series given input projection angles and 3D volume.

References

- [1] Rick Archibald, Anne Gelb, and Rodrigo B. Platte. Image reconstruction from undersampled Fourier data using the polynomial annihilation transform. *J. Sci. Comput.*, pages 1–21, 2015.
- [2] K. J. Batenburg and J. Sijbers. Dart: A practical reconstruction algorithm for discrete tomography. *IEEE Transactions on Image Processing*, 20(9):2542–2553, Sept 2011.
- [3] K.J. Batenburg, S. Bals, J. Sijbers, C. Kbel, P.A. Midgley, J.C. Hernandez, U. Kaiser, E.R. Encina, E.A. Coronado, and G. Van Tendeloo. 3d imaging of nanomaterials by discrete tomography. *Ultramicroscopy*, 109(6):730 – 740, 2009.
- [4] Tony Chan, Antonio Marquina, and Pep Mulet. High-order total variation-based image restoration. *SIAM J. Sci. Comput.*, 22(2):503–516, 2000.
- [5] B. Goris, T. Roelandts, K.J. Batenburg, H. Heidari Mezerji, and S. Bals. Advanced reconstruction algorithms for electron tomography: From comparison to combination. *Ultramicroscopy*, 127:40 – 47, 2013. Frontiers of Electron Microscopy in Materials Science.
- [6] Chengbo Li. *An efficient algorithm for total variation regularization with applications to the single pixel camera and compressive sensing*. PhD thesis, Rice University, 2009.
- [7] Chengbo Li, Wotao Yin, Hong Jiang, and Yin Zhang. An efficient augmented lagrangian method with applications to total variation minimization. *Comput. Optim. Appl.*, 56(3):507–530, 2013.
- [8] T. Sanders. Discrete iterative partial segmentation technique (dips) for tomographic reconstruction. *IEEE Trans. Comput. Imag.*, 2(1):71–82, March 2016.
- [9] Toby Sanders, Anne Gelb, Rodrigo B Platte, Ilke Arslan, and Kai Landskron. Recovering fine details from under-resolved electron tomography data using higher order total variation l1 regularization. *Ultramicroscopy*, 174:97–105, 2017.
- [10] Toby Sanders, Micah Prange, Cem Akatay, and Peter Binev. Physically motivated global alignment method for electron tomography. *Advanced Structural and Chemical Imaging*, 1(1):1–11, 2015.
- [11] Jeannot Trampert and Jean-Jacques Leveque. Simultaneous iterative reconstruction technique: Physical interpretation based on the generalized least squares solution. *J. Geophys. Res.*, 95(12):553–9, 1990.