

“Traces” analysis code for smFRET: Documentation

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1 Introduction.

Traces is software designed to calculate FRET-versus-time trajectories from a standard prism-based TIRF single-molecule FRET (smFRET) microscopy setup. The most recent version of Traces can be found at <<https://github.com/stephlj/Traces>>. It can be run as a stand-alone analysis suite, but it is written as modularly as possible in the hope that it can be adapted fairly easily to other microscopy setups and acquisition software.

I have attempted to avoid as much as possible “buried” parameters, collecting them instead in a TracesSetup.m file. TracesSetup.m is the only function a user should have to edit regularly (though some initial modifications may need to be made to, for example, the image loading function, as described below).

The general outline of the analysis workflow is: (1) Calculate a map that correlates pixels in the acceptor channel image to pixels in the donor channel image, as these will never be perfectly aligned in an smFRET setup. Or, load an old one. (2) Find fluorescent spots in a movie or set of movies, calculate intensity-versus-time traces for each spot, and then allow the user to save good traces for downstream analyses.

The rest of this manual contains information about required Matlab toolboxes, a walkthrough of how to run an analysis session off-the-shelf, and details of how the analysis is done.

1.1 Licensing and copyright information.

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1.2 Contact information.

If you find a bug in Traces, please open an issue through the Traces GitHub repository (<<https://github.com/stephlj/Traces>>). Please include as much information as you can about what causes the bug—what commands you entered, a screen shot of the output, etc. Thanks!

2 Getting started.

2.1 System requirements.

Requires Matlab (has been tested with R2012b through R2018b), with the following toolbox installed:

- Image Processing Toolbox

and will use the following if available:

- Curve Fitting Toolbox
- Optimization Toolbox

Disk space: The background subtraction routine (see Sec. 3.2.3) creates a set of additional files that get saved to the same directory as the raw image file(s) to be analyzed, and could be as large as 4 times the original movie size. Meaning, if you take a 1 GB movie, whatever drive you load this 1 GB movie from needs to have at least 4 GB of free space!

2.2 Data format requirements.

This software assumes the following about how raw data area acquired and stored from the camera:

- We acquire data with the open-source Micro-Manager software (<http://www.micro-manager.org/>), using its Multi-D acquisition tool and the “save as separate image files” option selected, which determines the data format requirements listed below.
- The function LoadUManagerTifs, called by LoadRawImg, assumes images are stored as “img*.tif”, with one tiff per frame.
- The function LoadRawImg assumes the file and directory structure created by Micro-Manager (summarized in Fig. 1); if you use something other than Micro-Manager to acquire data, and want to use this suite out-of-the-box, you will have to reorganize your data into the directory structure that LoadRawImg and Traces recognize.

Alternatively, you can modify LoadRawImg and its dependences (see Figs. 3 and 4 below), and some sections of the main Traces.m function, to recognize your data structure (those are probably the only two sets of functions you'll have to modify).

- Traces requires the metadata.txt file automatically created by Micro-Manager. If your acquisition software does not save a metadata.txt file, it is easy to create one yourself (alternatively, you can modify GetInfoFromMetaData.m in Traces/core). Each directory to be analyzed must have a metadata.txt.

In Matlab, change the current working directory to Traces/tools, and type “help GetInfoFromMetaData” into the Matlab command line for information about how to create metadata.txt files. A sample metadata.txt file is also included with this software, in Traces/SampleFiles.

- Alternatively, Traces can analyze the .pma binary files saved by the Ha lab acquisition software (and used by their IDL analysis software [<<https://cplc.illinois.edu/software/>>]), instead of tiffs. In Matlab,

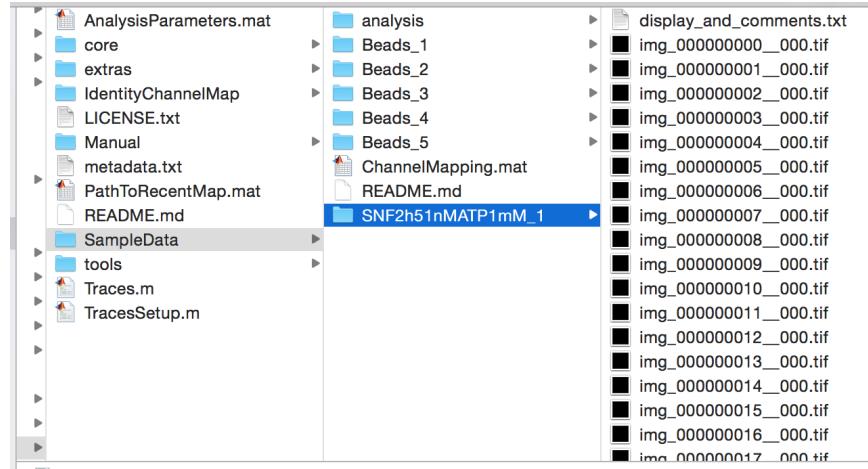


Figure 1: Screen shot illustrating the directory structure assumed by Traces.m and LoadRawImgs.m. Here “rootname” is “SNF2h51nMATP1mM”; running Traces(‘SNF2h51nMATP1mM’) will analyze the data in the folder selected in the middle column. If there were an additional directory called “SNF2h51nMATP1mM_2”, that directory would be automatically analyzed afterwards from the same call to Traces. If your acquisition software creates pma’s rather than tif’s, the “SNF2h51nMATP1mM_1” directory would have one .pma file instead of all of the .tif files, and a metadata.txt file, but not the display_and_comments.txt file (a file saved by Micro-Manager that we don’t use). **Note: Traces also supports newer versions of Micro-Manager that save the .tif image files, and the metadata file, in a subdirectory called “Pos0”, as long as Pos0 is the only image-containing subdirectory that you want to analyze.**

change the current working directory to Traces/tools, and type “help LoadPMA” in the command line in Matlab for more information about the assumed structure of these .pma files.

However, the directory structure created by Micro-Manager is still assumed—that is, you should put each .pma file in its own directory, and, if you want to analyze multiple pma’s in one batch, call each directory rootname* (e.g., rootname_1, rootname_2, etc) with each of these directories having only one pma. The filename of the pma’s can be whatever you want, but there can only be one per directory. NOTE: if both a .pma and a tiff(s) exist in the same folder, it will default to using the .pma!

- Channel mapping (at least an initial one) is accomplished by means of images of fluorescent beads that are visible in both channels. It is assumed these beads are in directories that contain the word “Bead” (e.g., “Beads_1”, “Beads_2”).

Although it is not assumed there are movies (rather than single frames) in these Bead directories, I have found it better to average 10-20 frames before finding bead positions. It is possible to skip the channel mapping with beads step (see below where the channel mapping procedure is discussed), but your channels will have to start out fairly well aligned.

- Some parameters that pertain to the microscopy setup you used to acquire images can be changed in TracesSetup.m, such as whether the donor channel is the right half or left half (or top or bottom) of the image. However, you will have to write your own wrapper function (Traces.m equivalent) for more complicated extensions, such as images obtained on a three-color FRET setup.

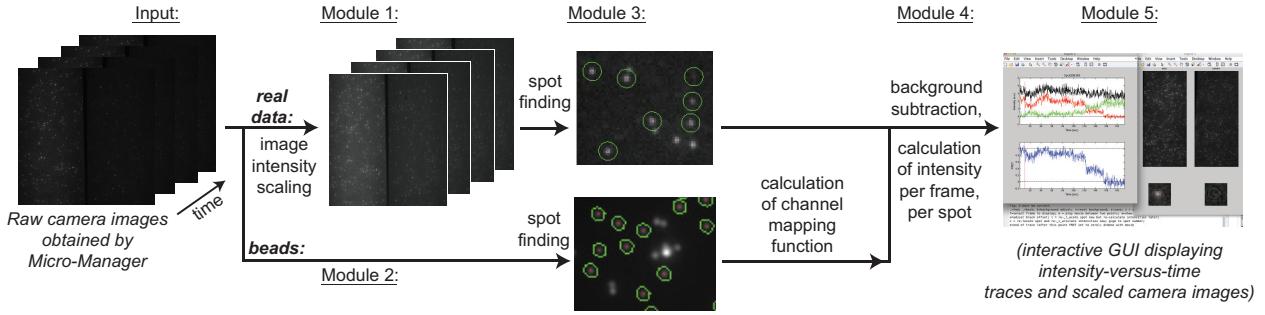


Figure 2: Schematic of the general Traces workflow. Note that some options, such as using paired spots selected in Module 5 as inputs to the channel mapping functionality of Module 2, are not shown. In these images the acceptor channel is on the left and the donor on the right. A screen shot of the interactive GUI is shown for Module 5; the two windows display (left) acceptor, donor, and acceptor+donor intensities (red, green, and black, respectively, top), FRET values (blue, bottom), and (right) acceptor and donor channel images of the spot that generated this trace.

2.3 Overview of the analysis process.

The workflow for analyzing smFRET data is, generally speaking, composed of five steps, which I call “modules” (Fig. 2):

- (1) data pre-processing, which includes, for example, scaling each image to take advantage of the full dynamic range of the numeric type you’re using;
- (2) finding a mapping between the donor and acceptor channels, which will never be perfectly aligned in any setup, such that a mathematical relationship between points in one channel to points in the other must be calculated to best identify FRET pairs;
- (3) spot finding, that is, identifying which pixels contain fluorescence from real molecules, and which are background fluctuations;
- (4) calculating the intensity in each identified spot in each frame, and thereby FRET-versus-time traces;
- (5) displaying these traces for the user to comb through to pick out real results for further analysis.

Note that modules are not necessarily performed in the order listed above; for example, we do a channel mapping (Module 2) before data pre-processing (Module 1), and this analysis suite also offers the opportunity to make a refined map (Module 2) using FRET pairs that the user has looked through (Module 5).

Details of how to run each module, and of what each module does, are discussed in the next sections. Figures 3 and 4 show call graphs that diagram the functions that perform each module, with required or optional Matlab toolboxes noted to the right of each function. Functions in colored boxes are “tools” that are or could be called by multiple functions; pink are highly specific to our microscope setup (and, along with the functions in white boxes, can be found in the “core” subdirectory), yellow should be more general (and can be found in the “tools” subdirectory). Blue asterisks indicate interactive sections requiring user input; red arrows indicate files saved to or loaded from disk (dashed lines indicate optional pathways based on user input).

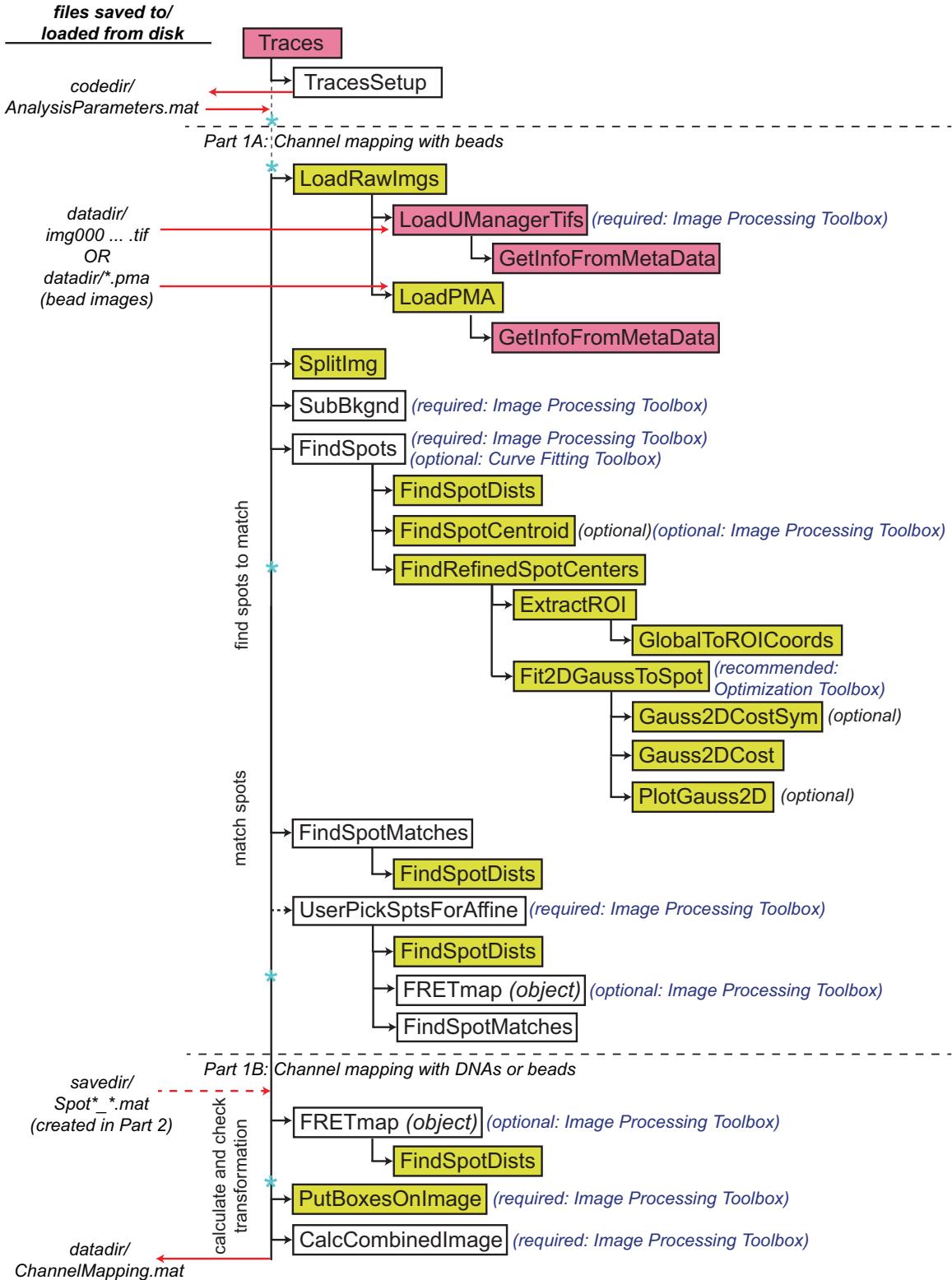


Figure 3: Call graph of the functions for calculating a channel mapping.

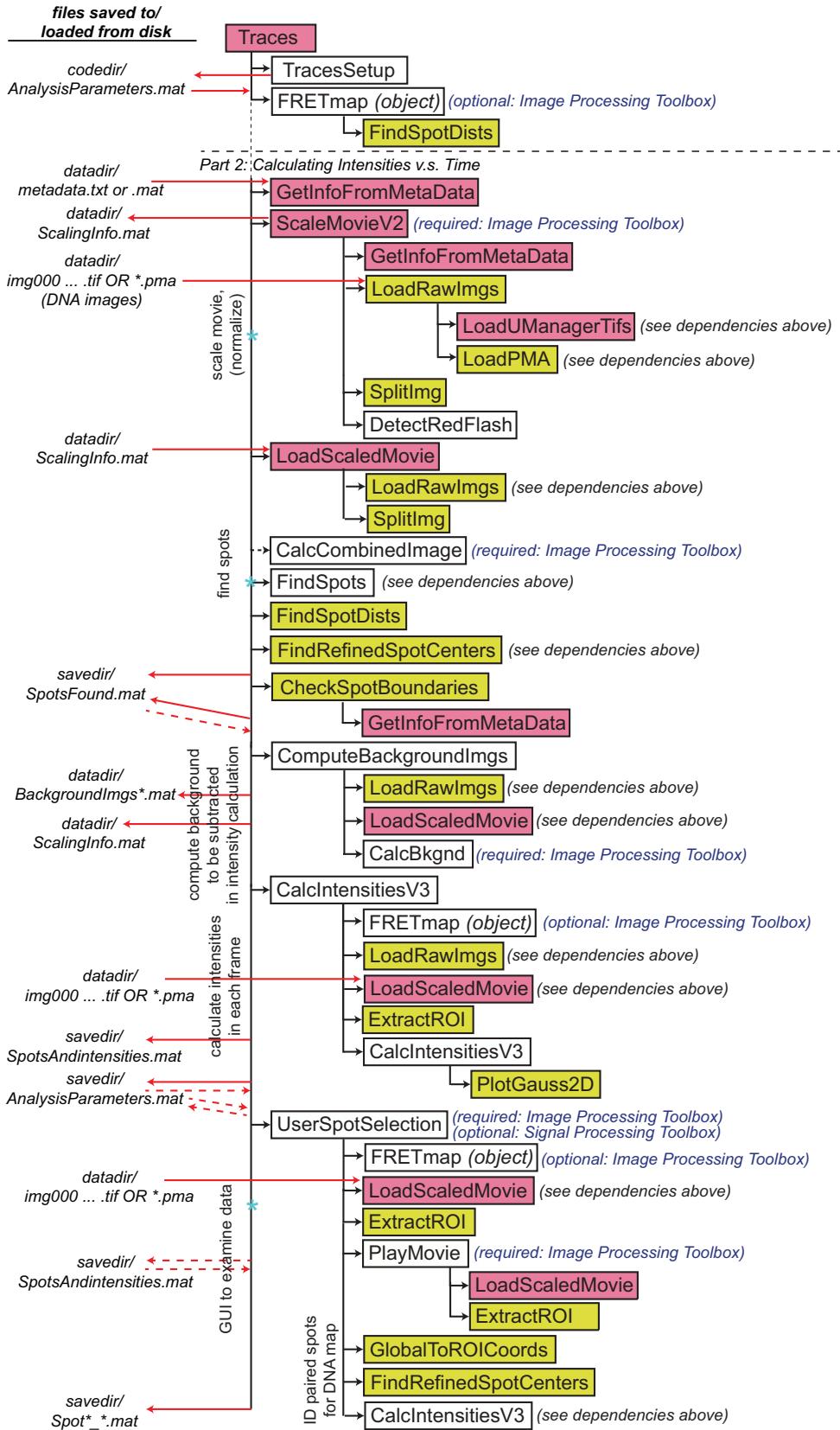


Figure 4: Call graph of the functions for analyzing real data.

2.4 Basic setup.

1. Copy the TracesSetup.m file in the SampleFiles directory to the same directory that contains Traces.m (should be the parent directory to SampleFiles).
2. Make the directory containing the Traces analysis suite code your current working directory in Matlab.
3. Open the TracesSetup.m file in Matlab and change the following parameters:

codedir: The full path to the directory containing the Traces.m file (and now TracesSetup.m).

defaultdatadir: Set this to the full path of the parent directory of your acquired images. If you're going to run Traces on the provided sample data (see Section 3), this should be whatever directory contains Beads_1, Beads_2, etc, and SNF2h51nMATP1mM_1.

defaultsavedir: Where Traces should save the analysis.

4. Follow the instructions under Display Defaults in TracesSetup.m to make sure figures will show up well on your screen.
5. If you're running Traces on the provided sample data, leave the rest of the parameters alone for now, and proceed as in the next section. If you want to adjust other parameters in TracesSetup, I suggest reading Section 4 first to understand the options.

NOTE: Any time Traces asks you to “press anything else”, it will probably crash if you press the space bar. Choose a different anything else! (Letter keys are usually safest.)

3 Using Traces to analyze provided sample data.

- (1) To start the analysis, enter

```
Traces('rootname')
```

in the command line (see Fig. 1 for a description of “rootname”). Use “SNF2h51nMATP1mM” as “rootname” to analyze the sample data that comes with this software.

- (2) You will be asked (in the command line) if you want to create a channel mapping using beads (Part 1A), load an old one (Part 2), or create a channel mapping from paired DNAs saved in Part 2 below (Part 1B).

3.1 Part 1A: Creating a channel map from fluorescent beads.

We (and others, e.g., [1]), routinely image a slide with surface-immobilized fluorescent beads to calculate at least a preliminary channel map, that is, a mathematical relationship between the donor and acceptor images that allows FRET pairs to be automatically identified. Generating such a map is part of a process called image registration; see Section 4.3 below for more details.

This software does not support the creation of a map from real data de novo, although there are workarounds that would allow it, as outlined in Section 3.5 below.

- (3) Press enter to perform channel mapping with beads.

- (4) Navigate to the directory that contains the bead directories.

If there is more than one directory with “Bead” in its title in the directory you choose, you will be given the option to use only some of the Bead data in creating a map, and to use the remaining bead data to quantify the fidelity of the map (always a good idea to keep some data “held-out” to check the transformation).

If you’re analyzing the provided sample data, I suggest answering 3 to “How many bead files to use for transformation”, and and answering “y” to using the remaining data to check the results.

3.1.1 Spot finding.

To calculate a channel map, spots will be identified in each channel based on a local thresholding algorithm (see Section 4.1 below), and then paired as in in the next section. A mathematical transformation from the spots in one channel to their paired spots in the other channel will then be calculated.

Traces will attempt to determine the correct threshold for finding spots, and will display an image of the green channel with circles around the spots it found. The number of spots (“peaks”) found is printed to the command line.

- (5) Press enter to accept the results, anything else+enter to choose a new threshold. A screenshot taken at this point in the analysis is shown in Fig. 5(A).

- (6) If you rejected the results, a histogram of the differences between all local maxima and local minima will be displayed (as in Fig. 5(B)), with a vertical black line at the current threshold value. For the beads, there should be two clear peaks in the histogram, one near 0 (which represents fluctuations

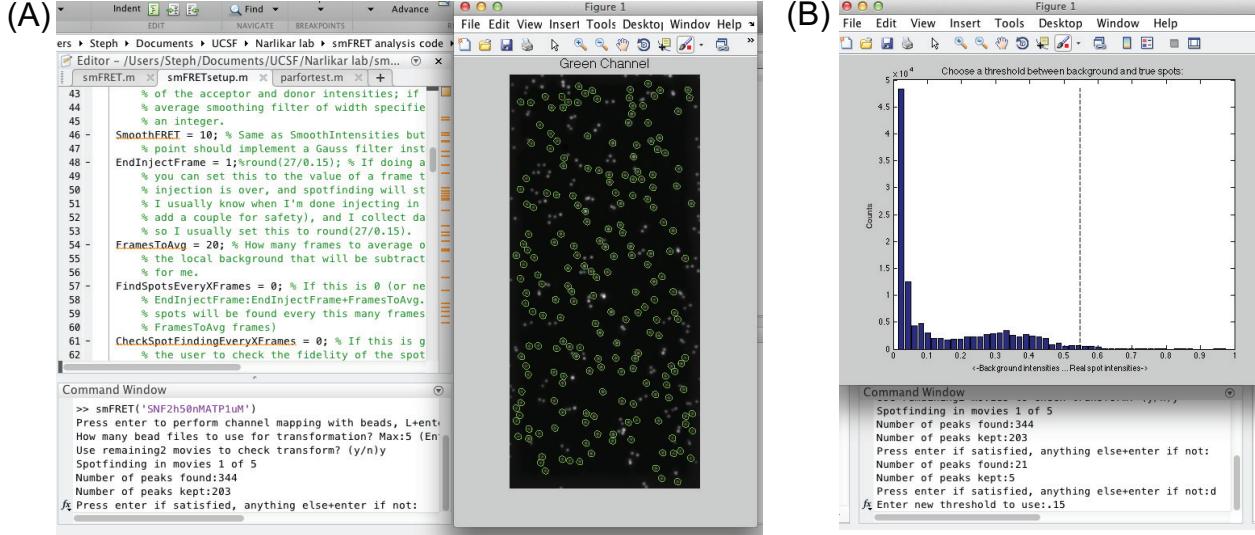


Figure 5: Screen shots of the spot finding routine. These are for spot finding on beads, but would be similar with real data. (A) The figure on the righthand side of the screen shows the donor channel half of the image from the camera, with green circles around each spot identified as a bead. (B) If the automatic threshold detection fails (as it obviously did here), you can set the threshold manually, using as a guide a histogram of the difference between each local maximum and its local minimum. With the beads it's easy to identify a good threshold—notice the two clearest peaks, one near zero that represents background, and one around 0.3 that represents the beads. The vertical dashed line indicates the current threshold value.

in background), and one which on our setup peaks around 0.2-0.3 (and maybe some small bumps at higher values that represent double beads, extra bright junk on the slide, etc). (These peaks won't necessarily be so clearly distinguishable with single dyes, but the same principle will apply there too.) Choose a new threshold value and press enter. Spot finding will be redone with this new threshold, after which you will again be given the option to accept the results or try another threshold.

NOTE(1): you must enter a threshold value before pressing enter, even if you've changed your mind and want to keep what it used the last time! The current threshold prints to the command line, if you want to re-enter it.

NOTE(2): The number of peaks found can be changed by changing the threshold (larger threshold will result in fewer peaks found); the number of peaks kept can be changed by changing the BeadSize parameter in TracesSetup (or, for spot finding with real data, the DNASize parameter), and/or the cen_tolerance and VarTolerance parameters at the beginning of FindRefinedSpotCenters.m, and/or the bknd_tolerance parameter near the beginning of FindSpots.m.

If you're analyzing the sample data, the number of peaks found per channel should be about 200, and the number of peaks kept about 150.

(7) Repeat for the red channel.

(8) Traces will next display a figure of both channels together, with white circles around all the spots found, and pause execution. Press any key to resume.

3.1.2 Spot pairing.

(9) Traces will next attempt to figure out which spots in one channel should be paired with which spots in the other channel. If it succeeds, it will print some information (including the number of spots paired) to the command line.

If your microscope's channel alignment is pretty good, the number of spots paired should be about 80%-90% of the spots found. For the sample data, maxDist should be less than 10, and the number of spots paired should be around 120.

(10) If your channels aren't aligned well on the microscope, Traces may not get enough correct spot pairs to get a good enough transformation to pick out the mismatches. In this case, the user will be asked to select three pairs of spots that are true matches (see Fig. 6(C)), from which an affine map will be calculated that will then be used to automatically match the rest of the spots. It's best to choose three spots that span as much of the image as possible.

If you're wondering, why 3?—Three is the minimum number of points necessary to define an affine transformation. More details about affine transformations can be found in Sec. 6 below.

NOTE: If you have a lot of spots in the field of view and/or the mapping is pretty far off from the identity transformation, the matching algorithm may still fail even with a user-generated affine map to help it. If that happens, you'll be asked to pick a different set of three pairs from which to create an affine map; if it still fails, the number of spots will be reduced (if you start with enough) a couple times in an attempt to get a good set of matches. (The problem with too many spots, if each spot is more than a couple pixels off from its pair in the other channel, is that the closest spot in the other channel may not be the correct match, and the greedy algorithm will fail.)

After spots have been paired, a new figure of the whole image will be displayed, this time with only the spots paired circled in green. A second image that overlays the locations of the spots found in one channel on those found in the other, with a blue line connecting the ones it thinks are matches, will also be displayed, as shown in Fig. 6(A). This second figure is useful for visualizing the alignment of the emission side optics of the microscope. Execution will again be paused until the user presses a button.

(11) Continue spot finding and spot pairing for each of the remaining bead movies.

3.1.3 Calculating the transformation that maps from one channel to the other.

The mapping from one channel to another is accomplished by a mathematical transformation that relates a point in one channel to a point in the other channel. See Section 4.3 below for more details.

(12) After spots have been found and paired in all Bead directories, a preliminary mapping will be calculated, and then refined by the removal of obvious outliers, which are mis-paired or poorly localized spots. (Even a handful of mismatches—*e.g.* 5-10 mismatched pairs out of almost 1000 pairs total—significantly reduce the accuracy of the calculated transformation!) After every cycle of refinement, the remaining number of spots that are being used to calculate the mapping, and the residuals for the current map, will be printed to the command line. These residuals are the sum of the squared differences between mapped spot locations and real spot locations. The refinement process will insist that the residuals per spot drop below the ResidTolerance value in TracesSetup; if it can't manage that, it will halt in the debugger and print some instructions to the command line about how to proceed.

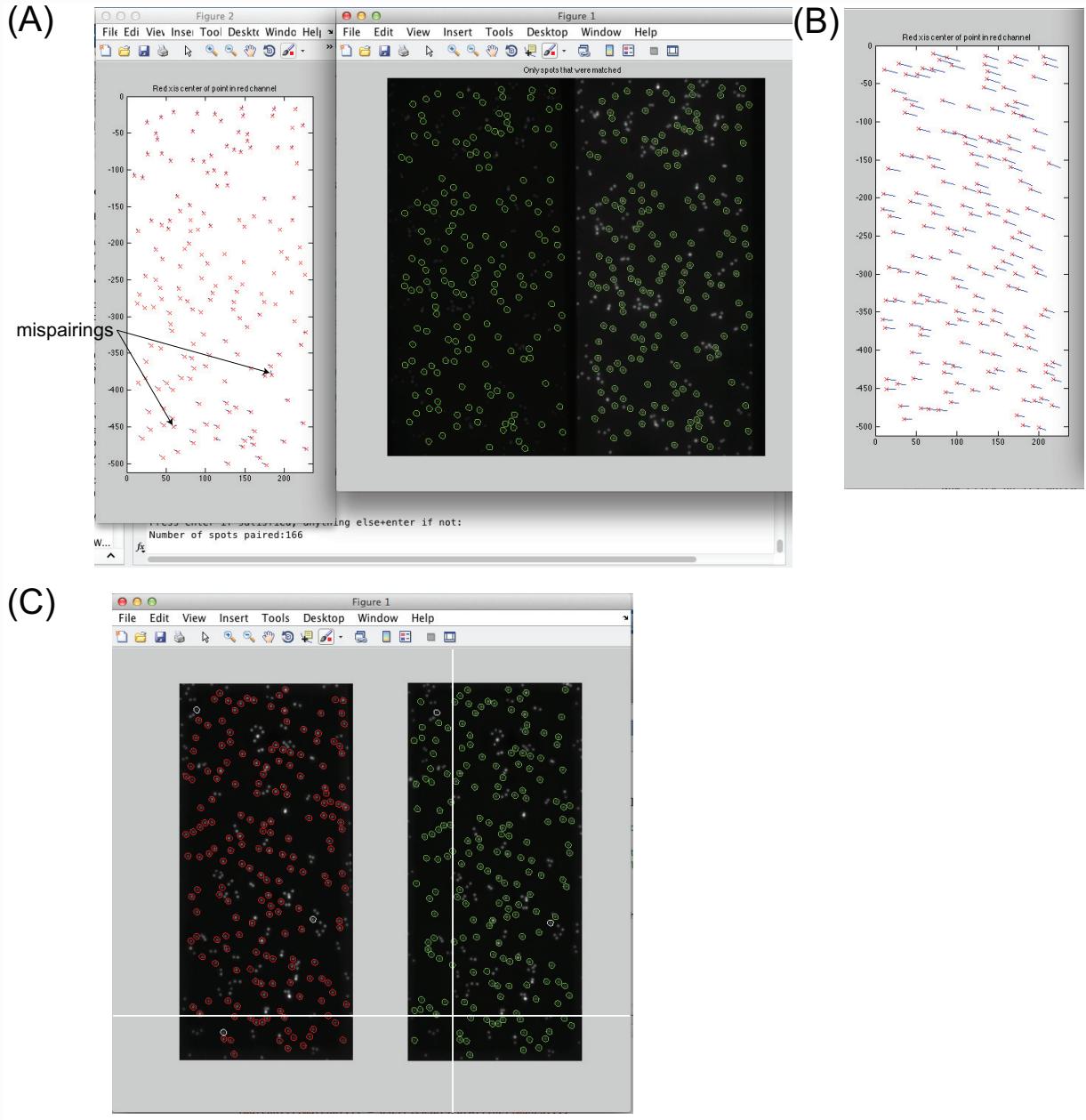


Figure 6: Screen shots of the figures displayed during the spot-matching routine, and of the manual pairing option for poorly aligned channels. (A) After spots have been found in each channel and then paired, a figure of the full image from the camera (left) with green circles around the paired spots will be displayed, and a plot of matched points in each channel overlaid on each other, with a blue line connected them (right). These figures are for a fairly good alignment on our microscope. Note the shape of the “vector field”—this gives a picture of the polynomial transformation that must be used to map between channels. Note also the mispairings—having only a couple mispairings, as here, is not a problem at this point; they will be removed before the final map is calculated, because they are obvious outliers. More than a couple mispairings, however, and it will not be clear enough what are outliers and what are real pairs for a map to be generated. (B) Same as the left figure in (A) but for a particularly bad alignment of the emission side optics. (C) For poor alignments like in (B), a greedy pairing algorithm will not work, since the correct spot to pair is not necessarily the closest one. In this case, the user will be shown a full image from the camera and will be asked to pick a spot in the red channel, then its pair in the green, then another spot in the red, until three pairs have been picked. (Specific instructions are printed to the command line.) Spots are circled in white after they are selected. White lines show the cross-hairs that Matlab displays for the user to pick points.

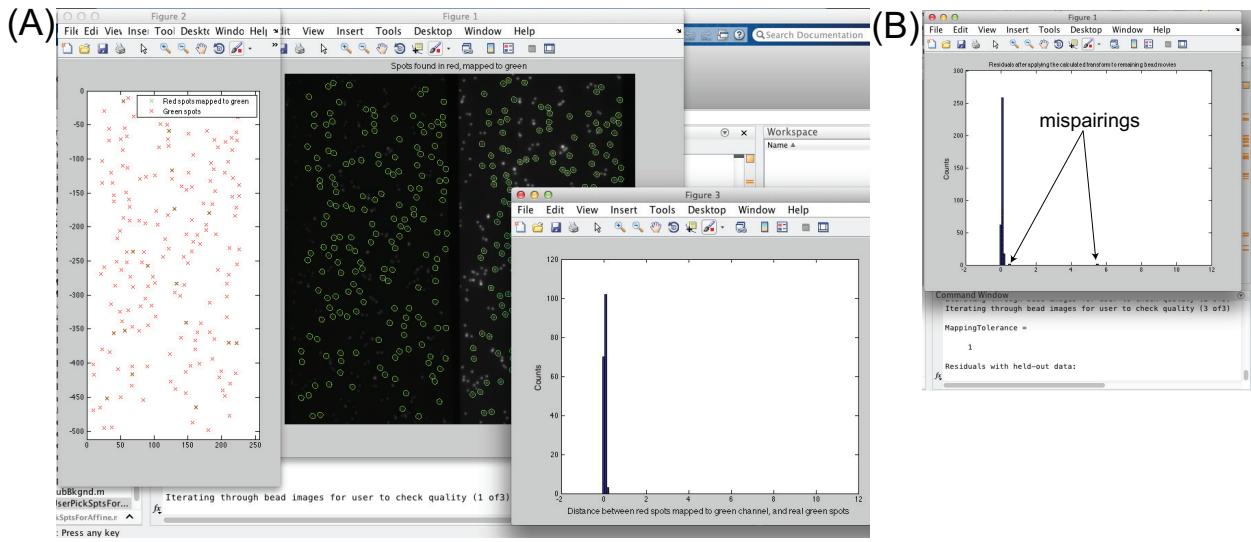


Figure 7: Screen shot of the figures that are displayed for the user to check the channel mapping. (A) For each Bead movie that was used to calculate the mapping, two sets of figures will be displayed, with a pause in between. The first set contains 5 figures that allow the user to check the mapping fidelity from donor to acceptor channel, three of which are shown here: an image of the full view of the camera, with the spots used to calculate the mapping circled in green; a scatter plot of the locations of the centers of all the acceptor-channel spots, and of where the mapping thinks their centers should be; a histogram of the distances between the real acceptor-channel spots and where the mapping thinks they should be; and overlays of the donor channel onto the acceptor, using either an affine or a polynomial transformation (not shown). (NOTE: if you have a version of Matlab older than 2013 or you’re using Traces’s custom transformation code, you’ll get a warning “CalcCombinedImage: your version of Matlab does not support this function”, and you will have only three figures, not five.) While the histogram is generally the most useful of these figures, the scatter plot and image can be useful for picking out where the mapping is failing, if it is. Note that there should be no counts above about 1 pixel in the histogram. The second set of figures is the same as the first three listed here, but for the inverse (acceptor to donor) transformation (the images shown here are actually for the inverse transformation). (B) If the user chose to keep some bead images as held-out data on which to check the transformation (that is, spot-find and pair in some images, but don’t use this information to calculate the transformation), after the previous sets of figures have been displayed for all bead images used to calculate the transformation, a histogram will be displayed showing the distances between mapped acceptor spots and real acceptor spots. In this case, there may be some residuals at higher pixel values, which should be mis-pairings in the held-out data. Only the donor-to-acceptor residuals are plotted.

Note that the residuals printed are summed over all spots; the ResidTolerance parameter is per spot, so a “ResidualsGtoRperSpot”, which is the residuals divided by the total number of spots included in the transformation, is also printed to the command line, for comparison to the tolerance. Note also that because we usually calculate a polynomial transformation, which is not invertible (unlike affine), a separate transformation is computed for acceptor to donor channel, and donor to acceptor; therefore both sets of residuals (“ResidualsGtoR” and “ResidualsRtoG”, where “R” stands for “red” and means “acceptor”, and “G” means donor (“green”)) are printed. There are also two sets of figures that display, one for one direction, one for the reverse. They’re generally pretty similar (both the figures and the residuals).

(13) The final channel mapping will be saved (along with some additional information) in the same directory as the Bead files used to generate it. The map is contained in a FRETmap object.

(14) A number of figures (see Fig. 7) are then displayed so that you can check how well the transformation did. The most important of these are the histograms of distances between mapped spots and real spots. For the held-out data, if any (that is, bead data that wasn’t used for the transformation), the vast majority of these distances should cluster between 0 and 0.5 pixels. For the non-held-out data, there should be no mismatches, and the differences between real spots and mapped spots should be mostly below ~ 0.2 pixels.

(15) Lastly, Traces will print a Mapping Tolerance to the command line. This is the maximum distance, rounded up to the nearest whole pixel, at which you can be confident a spot in one channel and a spot in the other are a pair. For a good channel mapping on our setup, the Mapping Tolerance should be 1 for beads (2 for single dyes).

NOTE (1): In order to (manually) load the FRETmap object back into Matlab, if you want to play around with it, you must have the directory containing the FRETmap.m class definition file in your known paths (or in your current directory).

NOTE (2): One of the sets of figures that are displayed at the end, if your version of Matlab is 2013 or later, includes overlays of one channel over the other, using either an affine transformation (which always does better) or polynomial (which always does worse). This would be used to calculate a combined image, if you wanted to spot-find using a combined image to more easily catch mid-FRET spots, as the Ha lab code does. For reasons that I’m still not clear on, it (very obviously) does not work with Matlab’s built-in overlay functionality ... so for the moment, at least on our setup, using a combined image to find spots is not an option. I’ve not had a problem, however, finding mid-FRET spots, by changing the threshold in the spot-finding routine, and/or by spot-finding every 100 or so frames.

3.2 Part 2: Generate and examine FRET-versus-time traces.

After you’ve calculated a map, or have loaded an old one, navigate (when prompted) to the directory that has the data directory/directories you want to analyze, whose names should start with the “rootname” entered as the argument to Traces.

3.2.1 Scaling and (optionally) normalizing images.

Scaling refers to taking the raw images from the camera, which are sometimes in a slightly weird numeric type (like uint14’s), and whose actual intensity values don’t span the full available range of that

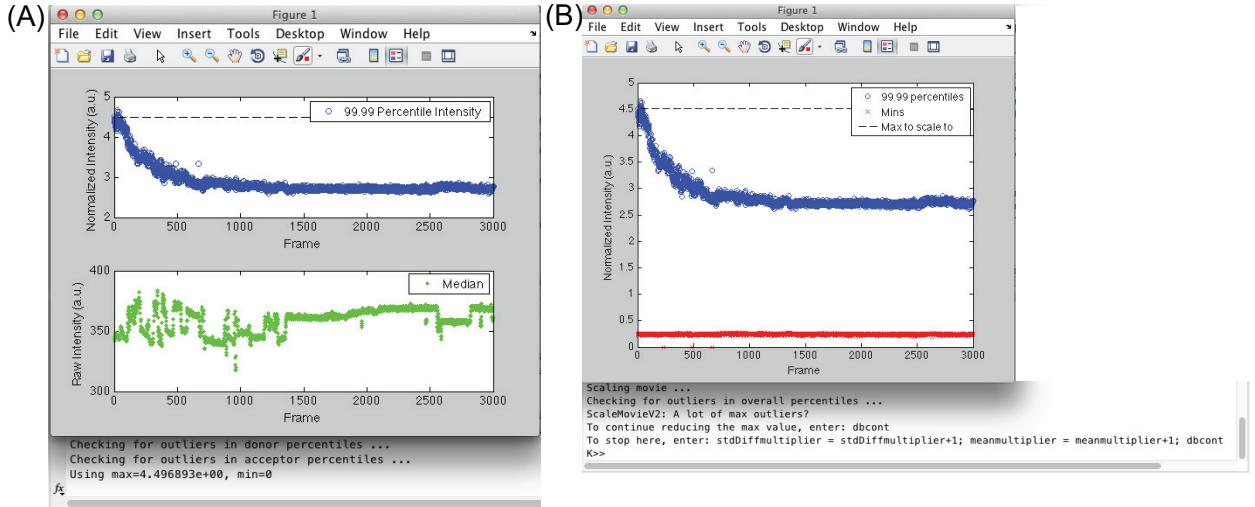


Figure 8: Screenshots for the movie scaling (data pre-processing) steps. (A) For every frame in a movie, the 99.99th percentile (top panel, blue) and median (bottom panel, green) intensity values are calculated and displayed. The horizontal black line on the top panel indicates the max value to which all intensities will be scaled. The maximum and minimum values to be used for scaling are printed to the command line and execution is paused so that the user can confirm the values are reasonable. (Note that if `ScaleChannelsSeparately` is set to 1 in `TracesSetup`, there will be an additional panel between the two shown here, that shows the 99.99th percentiles for the two channels separately, and the max and min to which each channel will be scaled will be printed to the command line.) The bottom panel shows the roughly 7% median intensity value fluctuations that we have observed with an old laser or when extraneous light was not sufficiently blocked from the emission optics (and camera). If these are not removed, these fluctuations will show up in all intensity-versus-time traces for individual spots and obscure real signal. If the `NormImage` parameter in `TracesSetup` is set to 1, each frame will be normalized to (specifically, divided by) its median value, removing the median intensity fluctuations. Obviously, though, it is better to eliminate these fluctuations by changing your microscope than to do so computationally, when possible. (B) To choose a reasonable 99.99th percentile to use as the max, outliers in these percentiles are first removed. If more than 7 of them are removed, execution will pause and ask the user to confirm that the scaling algorithm is truly removing outliers or if it is starting to remove real maxima. If the maxima plot looks like it does in (B), where there are more than 7 clear outliers, follow the instructions in the command line to override the checkpoint. (You may have to do that several times, if there are more than 8 outliers in the maximum intensities, because the scaling algorithm will pause after each one to check that it's still only removing outliers.)

numeric type, and doing two things to them: converting to a more common numeric type (like uint8’s, or doubles), and linearly expanding the image intensities to take advantage of the full dynamic range available. More information can be found in Section 4.4.

(16) Traces will display the “maximum” (99.99th percentile) and median intensity values for the movie as a function of frame, and print the values it’s going to set to 0 (min across the movie) and 1 (max across the movie) to the command line (see Fig. 8(A)). Then it will pause and wait for you to press any button to accept and proceed.

If ScaleChannelsSeparately=1 in TracesSetup, it will also display the maximum intensity values for the two channels separately in a middle panel.

If it has to remove a lot of outliers to find a good maximum to scale to, it will stop in the debugger and ask for user input. Follow the instructions that print to the command line to proceed.

(17) Traces also provides the option of detecting a red laser flash, which we use as a fiducial mark for injection by our automated syringe pump. If the “DetectRedFlash” parameter is set to a value greater than 0 in TracesSetup.m, Traces will look for “flashes” of high intensity in the acceptor channel that aren’t in the donor channel. If such flashes are found (and confirmed by the user), an additional parameter called “InjectPoints” will be saved to the AnalysisParameters.mat file, containing the time index of these injection marks. They will be displayed in the GUI, with the InjectDelay (in TracesSetup) added, as a vertical magenta line.

(18) As described in Fig. 8, we’ve had problems with laser intensity fluctuations with one of our older lasers; therefore if NormImage is set to 1 in TracesSetup, all pixels in each frame will be normalized to the median intensity value in each frame to remove these fluctuations. This is obviously a sub-optimal way to remove fluctuations—it’s always better to remove them experimentally rather than computationally!—but it can be useful in the short-term while troubleshooting an smFRET setup.

3.2.2 Spot finding.

(19) Spotfinding proceeds as described for the beads in Sec. 3.1.1 above, with the option of the following modifications that are controlled by parameters in TracesSetup:

(a) If instead of only finding spots in an averaged image of the first FramesToAvg frames, you want to find spots throughout the movie, set the FindSpotsEveryXFrames parameter to a value greater than 0. After an initial round of asking the user to set a good threshold for real spots, spot finding will proceed with this same threshold until a multiple of CheckSpotFindingEveryXFrames (see next point) is reached.

(b) I highly recommend setting the CheckSpotFindingEveryXFrames parameter to a value equal to FindSpotsEveryXFrames or to 5*FindSpotsEveryXFrames, or something like that. At least with my data, the optimal threshold varies enough over, say, 100 frames that if you just keep using the threshold for the first FramesToAvg frames, you run the risk of a significant number of false positives (background fluctuations that get counted as real spots) or missing real spots because the threshold was too high or too low.

(20) At the end of the spot finding session, all spots found in both channels will be saved to the directory you chose to save data in (*savedir* in TracesSetup), in a file called SpotsFound.mat. In addition, a list of spots in the frame of reference of the acceptor channel will also be saved, as follows:

- Spots found in the donor channel are converted to the frame of reference of the acceptor channel, using the channel map that you calculated or loaded.
- If a donor spot was found close by to an acceptor spot (*e.g.*, if that spot was a mid-FRET spot and so showed up in both channels), the location of the spot in the acceptor channel will be saved—so that spots aren’t double-counted.
- This final list of spots in the frame of reference of the acceptor channel will be saved as the *spots* variable, and it is this list that will be used to calculate intensities for each spot.
- *SpotsFound.mat* will also contain *Vars*, *VarsR* and *VarsG*. These contain the *x*- and *y*-variances of the best-fit 2-D Gaussian used to refine each spot’s location. *VarsR* and *VarsG* are the variances for spots found in the acceptor channel (saved to the *RefinedCentersR* variable) and in the donor channel (*RefinedCentersG*) respectively. *Vars* has the same length as *spots* and contains the variances that will be used in the intensity calculation. For any spot found in only one channel, that spot’s variances is used; however if a spot was found in both channels, the variance of the spot in the acceptor channel is saved. These variances are used if the *IntensityGaussWeight* parameter in *TracesSetup* is 1 and *FixSpotVar* is an empty matrix.

Any time after this point, you can quit out of Traces, rerun it with the same input argument (*i.e.* the same rootname), load a different channel map, and then load the spots found in both channels and re-convert to the acceptor channel frame of reference, by selecting the “Load previously found spots?” option (NOT the “Load previously found spots and their intensities?” option). Spot intensities will then be re-calculated. (Since spot-finding is independent of a channel map and happens for each channel separately, you don’t have to find spots again just because you want to change where their pair in the other channel should be. There are also options to re-scale the movie without re-finding spots. See more details in Sec. 3.4.)

3.2.3 Calculating the background.

At this point a background image is calculated for each frame and saved to disk, for use in the next step when the spot’s intensity is calculated. See Section 4.5 for details. No user input is required.

NOTE: This is currently the longest step of analysis—on my laptop this takes ~5-7 minutes for a data set with ~3000 frames. Therefore this step is done only right before you need it—so that you can find spots and examine the quality of the movie in the previous step before investing the time to calculate the background.

Note also that because these background images are saved as double precision, for best compatibility with Matlab’s toolboxes, and there’s one for every frame of your movie, they’re going to take up a relatively large amount of additional disk space—for example, as much as 4x the original movie size, if the original movie was saved in int16 precision like ours are. You can delete these files after you’ve completed the next step (calculating intensities vs time), but they will be re-generated every time you analyze a particular data set.

3.2.4 Calculating intensities-per-time for each spot.

(21) Raw (that is, unsmoothed, if *TracesSetup* has smoothing options selected) intensities for donor and acceptor channels will be calculated and the results will be saved to the save directory, in a “*SpotsAndIntensities.mat*” file.

Raw data also does not have any corrections to the intensities (like alpha) applied.

Any time after this point, you can quit out of Traces and re-load traces into the GUI, without re-calculating their *raw* intensities, by selecting the “Load previously found spots and their intensities?” option. Use this option if, for example, you want to change the smoothing parameters. Change the relevant parameter(s) in TracesSetup, and then re-run Traces. When asked if you want to load the GUI with old parameters, select “n”.

3.2.5 The GUI.

(22) The GUI has:

- A figure showing intensity-versus-time and FRET-versus-time traces
- A separate figure showing (initially) an average of the first FramesToAvg frames for each channel, plus a zoomed image of the current spot in each channel below the full images.

The displayed intensities and FRET values are smoothed and have the correction factors gamma and alpha applied (see TracesSetup for definitions of gamma and alpha); that is, they are NOT the raw intensities (though these raw intensities are saved along with modified intensities—see below).

(23) Instructions for GUI options print to the command line. Most options should be straightforward, with the possible exceptions of:

“b” (background adjustment): If the background subtraction routine didn’t remove all the background (*i.e.*, intensity values are significantly nonzero after photobleaching), select this option, then click on two parts of the intensity-versus-time or FRET-versus-time trace over which you want to define a new background value. The average donor and acceptor intensities between these two points will be subtracted from every point. *Note that right now you can only subtract background from both acceptor and donor channels together!—so you must click two points that span a part of the trace where both dyes have bleached or blinked.*

“r” (reset background): remove any additional background subtraction you added with the “b” option.

“z” (zoom): Click on two points in the intensity-versus-time or FRET-versus-time traces to zoom in on that time segment. *NOTE that this will define how much of each trace is saved—see below.*

“u” (unzoom): undo any zoom.

“f” (show a particular frame), “m” (play movie between two time points), “a” (show average around a particular frame): In the case of “f” or “a”, click on a point in the intensity-versus-time or FRET-versus-time traces for which you want the image displayed in Fig. 1. Note that if you go forward or back a spot, Fig. 1 will reset to an average of the first FramesToAvg frames. “m” will leave Fig. 1 showing the last frame loaded (but again Fig. 1 will reset if you go forwards or backwards).

“c” (re-locate spot and re-calculate intensities): If the channel mapping didn’t find a spot pair well in one channel, you can click near where you think the true spot center is, and Traces will attempt to re-find the spot in that area, and then re-calculate the intensities using that new spot center. (This may take a few minutes.) Note: if a new spot center is successfully found, this new center, and the newly calculated intensities, will be saved to the “SpotsAndIntensities” file. If you want to revert to original spot locations, you’ll need to re-load from the “Load previously found spots?” point.

“l” (re-locate spot): The same as “c” but the spot is only re-located immediately; it is then added to a list of spots whose intensities are re-calculated all at once at the end. The user is then given

the opportunity to page through these re-located spots, with all the usual GUI options except this “l” option (“c” is still available). Also, information about spots that were re-located but whose intensities haven’t been re-calculated are saved to the “SpotsAndIntensities” file, so the user can quit out of the analysis and pick up where they left off later, without losing this information. This option is useful if you have a lot of spots to re-map—since the slowest part of this computation is loading from disk, it’s much, much faster to load the movie once and re-calculate lots of spots’ intensities at once, rather than doing one spot at a time.

You need to make all selections with Fig. 2 current—if you don’t, you’ll notice you’ll just type stuff into the command line and nothing will happen ...

DON’T press enter after making a selection—for example, to go forward one spot, just press “.”, no enter needed. Any time you enter something into the command line, however, you do need to press enter—for example, if you select “g” to go to a particular spot, you will be prompted to enter the spot number in the command line, after which you *will* need to press enter.

(24) What is saved to disk when you select the “s” option in the GUI: When you select “s” for a spot, a new file will be created in the save directory called “Spot<movienum>_<spotnum>.mat” (or if you select “s” for a spot for which a file has already been created, you will overwrite the previous file with a new one). This file contains:

- (a) vectors of raw donor and acceptor channel intensities for each FRAME (“rawRedI”, “rawGrI”);
- (b) vectors of unsmoothed, but with alpha applied for the acceptor intensities and gamma applied for FRET values, donor and acceptor intensities and FRET values for each frame (“unsmoothedRedI”, “unsmoothedGrI”, “unsmoothedFRET”). *It is these intensities and FRET values that are used in downstream analyses like vbFRET [2] or the HMM analysis included with Traces (see Section 5 below);*
- (c) vectors of donor and acceptor intensities and FRET values with any smoothing applied (“RedI”, “GrI”, “FRET”—these will be identical to the unsmoothed versions if no smoothing was applied;
- (d) The frame rate at which data were acquired (frames per second, “fps”);
- (e) The (x, y) coordinates of the center of the spot in the donor channel and the acceptor channel, in the coordinate systems of their respective channels (note these coordinates will depend on the value of PxlsToExclude in TracesSetup) (“Rspot”, “Gspot”), and the variance used for the Gaussian weighted sum, which is assumed to be the same for both channels (“variance”).

In addition to this Spot file, an eps and a Matlab .fig file of whatever is in Fig. 2 will also be saved, with the same filename. NOTE: Only what is in Fig. 2 will be saved—meaning if you zoomed in on part of the trace, the Spot file will only contain the intensities and FRET values for the zoomed region. You can choose to cut out parts of traces this way (for example, you can choose not to keep everything after both fluorophores have bleached). If, however, you decide you want to restore part of the trace you cut, just re-load this data set into the GUI and use the “u” command to un-zoom—the full raw intensities are still saved in the “SpotsAndIntensities.mat” file and will be re-calculated on the fly.

3.3 Part 1B: Creating a channel map from paired DNAs.

You can make a map using true paired DNAs (that you’ve confirmed are true FRET pairs by their intensity traces, and that you made sure, using the GUI, were well-localized) by running Traces with any input, selecting the “D” option, and navigating to a save directory that contains the “Spot<>.mat”

files saved with the “s” option from the GUI. As with the beads, a transformation will be calculated, the residuals refined as much as possible, and then the user will be asked to check the results. This works the same as described above for beads, although most of the figures that are generated for bead maps will not be generated here. NOTE: You need at least 15 pairs of spots in order to calculate a polynomial transformation (at least of degree 4, as we and others usually do); if Matlab spits out a bunch of warnings about poorly conditioned fits, and/or you have close to or fewer than 15 spots, try changing to “MatlabAffine” in TracesSetup.

3.4 Options for re-analysis.

If you want to go back and re-analyze some aspect of a data set you’ve already finished, you have the following options after either re-making or loading a channel map:

⇒ If a “*SpotsAndIntensities.mat*” file exists from a previous analysis: “Load previously found spots and their intensities? (y/n)”

⇒ If you select “y”: “Use old parameters? (y/n)”, and go straight to GUI.

⇒ If you select “n” and a “*SpotsFound.mat*” file exists: “Load previously found spots? (y/n)”

⇒ If you select “y” (use old spots): “Use same channel mapping as last time (y) but recompute scaling, background, intensities? Or use new channel mapping (n), keep old scaling and background, but recompute intensities?”

If you select “y” (use old mapping): The same spots as were displayed in the GUI last time will be kept (including their numbering), but the movie will be re-scaled, and new background images and intensities for each spot will be computed.

If you select “n”: Spots found in the two channels will be loaded, and a new list of spots will be generated based on whatever channel mapping you loaded at the start. Use this option if you don’t want to re-find spots but do want to change the mapping you used to associate a red spot with a green spot. HOWEVER: note that this will change the numbering (and likely the total number) of the spots found, so any spots you found before will not have the same indices as the last time you ran the GUI. Spot intensities are recalculated, but not background images.

⇒ If you select “n” and a *ScalingInfo.mat* file exists: “Load scaled movie? (y/n)”

If you select “n”: the movie is re-scaled, then new spots are found, background subtraction information created, and spot intensities re-calculated.

If you select “y”: the movie is re-scaled and background is NOT recalculated, but new spots are found and their intensities calculated.

Note: To re-calculate the background without re-scaling, delete the “BackgroundImg.mat” files in savedir.*

3.5 Other analysis options.

Generating a channel mapping from real data de novo. Instead of using images of stuck beads to generate a channel map, you can load the FRETmap object included in this suite that contains the identity transformation instead of a calculated affine or polynomial transformation (“ChannelMapping.mat” in the SampleFiles subdirectory). Load this map using the “L” option, then use the GUI to generate paired spots that could be re-loaded using the “D” option. However, without a bead-based map to get you close enough to see the real spot pairs in the GUI, your channels would have to be fairly well aligned to start with in order to make this work.

4 What Traces does at each step of the analysis.

4.1 Spot finding.

Spot finding is accomplished by finding local maxima and local minima across the image, and then calling real spots those maxima that are sufficiently larger than their local minima; the rest of the local maxima, that fall below the threshold, represent fluctuations in background. This method has the advantage of not setting an absolute intensity threshold for real spots that must apply across the entire image, thereby allowing the illumination to vary across the image.

The pioneering Ha lab code (<<https://cplc.illinois.edu/software/>>) does something roughly equivalent: they use a single absolute threshold value across the entire image, but they spot-find on a background-subtracted image. (In the Ha lab code, the threshold is hard-coded into nxgn1.cm (spot finding for beads) and p_nxgn1_ffp and ..._brief (spot finding for DNAs), in the parameter “std”.) I prefer to keep spot-finding and background subtraction as separate procedures that can be modified independently without changes to one impacting the other.

Spot finding is done on one channel at a time, on an averaged image of the first FramesToAvg frames (where FramesToAvg is set in TracesSetup; the Ha lab uses a hard-coded value of 10 frames). The spot-finding routine in FindSpots will attempt to identify an appropriate threshold value that distinguishes background fluctuations from real spots. The number of spots (“peaks”) that, according to this threshold, are considered to be real signal are printed to the command line. A further refinement of all peaks found is then performed; the basic refinement that is always performed excludes spots too close to the edges or to another spot (where “too close” is defined as closer than DNASize or BeadSize pixels in TracesSetup). The Traces wrapper function passes an input pair (“Method”, “GaussFit”) that does an additional refinement by fitting a 2D Gaussian to each peak and insisting that the variance is not too small (which is usually the case when a very bright single pixel got included as a “spot”), and the local background is not too high (which happens when two beads are right next to each other, but only one was found as a local maximum). The number of peaks kept after these refinements are then printed to the command line.

I don’t think the Ha lab code excludes peaks too close to one another—though it does exclude spots that aren’t surrounded by sufficiently low intensity values at a radius of 5 pixels from the local maximum—so my code may find fewer spots per field of view than you’re used to if you use the Ha lab code, though hopefully a higher percentage of the spots found will be useful, especially for real data.

4.2 Spot pairing.

The pairing code uses a “greedy algorithm” to match spots, which essentially means it assumes for each spot that its match in the other channel isn’t too far away, and is probably the closest one. Some mismatches at this point are ok; they will be removed before the final mapping is calculated (even a couple mis-pairings out of hundreds of true pairs can really mess up the fidelity of the mapping!).

The Ha lab code does something more like what Traces does if the greedy algorithm fails: it asks the user to identify three pairs of spots, creates a rough affine map to match more spots, and then uses those additional spots to generate a polynomial transformation (which requires a minimum of 15 spots, for the polynomial used in our code, or minimally 16 spots for the polynomial used in the Ha lab IDL code). But, the more spots the better, because then you will have more information about the form the polynomial mapping takes for more of the field of view.

4.3 Channel mapping (image registration).

Not all smFRET software packages rely on the calculation of a channel map to identify FRET pairs; for example, iSMS (<<http://isms.au.dk/>>) allows the user to align channel images somewhat by translating them horizontally and vertically, asks the user to set thresholds that identify most or all spots in each channel, and then assumes that a spot’s FRET pair is whichever spot in the other channel is closest. This has certain advantages, including not requiring the use images of beads as a starting point. However, it works only when the spot density is relatively low, limiting the potential through-put of the assay, and really only works robustly if you use alternating laser illumination in the first part of your movie to illuminate all spots in both channels.

On the other hand, it is relatively straightforward to generate a channel mapping from immobilized fluorescent beads (we use Invitrogen TetraSpeck beads). A best-fit transformation is calculated from paired sets of points, one set in each channel. As discussed in more detail below (see Sec. 6), we and others have found a 4th-degree polynomial to best describe the relationship between channels. The degree and kind of polynomial fitted, and the option to use affine instead of polynomial, are set in TracesSetup. Both IDL and Matlab have built-in functions to fit polynomial or affine transformations; the FRETmap class object also includes hand-written code to do the same, which might be useful if, for example, you want to fit a maximum degree 4 polynomial, which Matlab’s built-in polynomial class does not support; see also Sec. 6.

4.4 Scaling and normalizing images.

Scaling means taking advantage of the full dynamic range available, which means telling Matlab what value to call “white” and what value to call “black”. Ideally you want your brightest pixels (those with the highest intensity values) to display as white, and the darkest (those with the lowest intensity values) to display as black. Let’s say your camera saves uint14 images; that means the available range of values each pixel can have is [0 16383]. Matlab, or any other program you use to display the raw images, will assume 16383 means white, and 0 means black. Usually, however, most pixel intensity values will be much, much smaller than 16383, so everything will display as essentially black.

There are three reasons you’d want to scale images. The first has to do only with presenting the data to the user in a helpful way, NOT with anything that actually affects the donor and acceptor intensity values, or FRET values, you get out of the data. By scaling the raw intensity values to the full dynamic range, the user can see all the features of the image clearly when the image is displayed on the screen, because you tell Matlab to pick a more reasonable intensity value than the never-approached

16383 to call “white”. As long as you do a linear scaling, scaling to the available range does not change anything about the actual information you get out of the images, because each pixel’s relative value to every other pixel remains the same. The second reason you’d want to scale your images is related to the first, in that many of Matlab’s built-in image processing toolbox functions want images scaled on the interval [0 1] (which is, for our purposes here, Matlab’s preferred “dynamic range”). That is, Matlab wants the brightest intensity value, the one you want displayed as white, to have a value of 1, the lowest intensity value, which should display as black, to be zero, and everything else scaled between 0 and 1. Many of Matlab’s functions also want to be given doubles, not uint16’s or uint8’s; the built-in Matlab function that scales a given image or set of images to [0 1], `mat2gray`, also converts to double precision. Finally, depending on your setup, it may be necessary to scale the donor and acceptor images to different maxima—meaning, the value that gets set to “white” for acceptor may not be the same as for donor images—so that acceptor and donor intensities can be directly compared (discussed more below). Unlike scaling the entire (donor+acceptor together) images to a maximum that makes displaying the image nice, scaling the donor and acceptor images separately will, obviously, change your results (most importantly, your calculated FRET values).

Choosing the maximum value to which all intensities should be scaled can be nontrivial. The Ha lab acquisition code scales to minimum and maximum intensity values set (with some user oversight) before acquisition begins. The UCSF mrc to pma converter scales to the minimum and maximum intensity values across the entire movie. This can be sub-optimal, because there are often a few very, very bright pixels (outliers in the distribution of intensities, often caused by “dirt” on the slide or similar) that will get set to white and everything that you care about set to nearly black. Since the UCSF mrc to pma converter is only used to make our data compatible with the Ha lab IDL code, which doesn’t display images for the user to inspect anyway, the maximum the converter chooses doesn’t really matter. However, since Traces does frequently display images for the user, I have found that a better way to scale each movie is to use an intensity value that represents the 99.99th percentile of all the intensities in the movie.¹ It’s very important to make sure that no real data intensities get set to 1, the highest possible value, so that you don’t “cap” real information in your image, which is why the 99.99th percentile is chosen (that should be well above the actual intensities of any real data, well out into the tail of very-bright-dirt outliers in the intensity distribution). ScaleMovie will display intensity histograms for the user to check that the maximum value, to be set to 1, is indeed well out into the tail.

More importantly for our setup, though, is the option of scaling the donor and acceptor images to different maxima. In TracesSetup you can set whether to scale the donor and acceptor channels separately, using the `ScaleChannelsSeparately` parameter. Possibly because of some of the filters in our microscope setup, our donor channel is always slightly dimmer than the acceptor channel; therefore we scale the donor images to the 99.99th percentile value that the donor side takes on, and the acceptor images to the 99.99th percentile value that the acceptor side takes on. This essentially puts the donor and acceptor intensities reported for each spot in the GUI on the same set of axes, so that they can be directly compared to one another.

You can tell that a movie is appropriately scaled by two metrics: first, that the acceptor and donor spots are easy to see in the images that display with the GUI (see below); and second, that the maximum acceptor and donor intensities in every trace are roughly equal. That is, if you have a FRET-ing molecule, and the acceptor bleaches, the donor should rise to the same value as the acceptor (unless you have reason to believe that shouldn’t be the case for your system).

¹Technically, ScaleMovie calculates the 99.99th percentile from each frame, and then takes the max value of those percentiles. It works out to be about the same, and doesn’t much matter anyway.

4.5 Background subtraction.

A visual inspection of TIRFM images will demonstrate that the background is not constant over the entire field of view (to a greater or lesser extent depending on the shape of the TIRF spot relative to the field of view, among other things). Moreover, we know that every image will consist of a lot of background, with some very localized real signal (that is, spots that represent fluorescence from the dyes). Our background estimation routine first smooths the images, to smooth out contributions from the punctate real data, and then median filters² the image to pick out the slowly spatially-varying background (and then smooths this calculated background a little more). This background calculation procedure is essentially the same as the Ha lab's, except we calculate one background image for every frame instead of one for the whole movie from an average of the first 10 frames, allowing for a time-varying background due to laser intensity fluctuations or other factors.

4.6 Calculating intensities per spot per frame.

Nominally, calculating the intensity of a spot in each frame is simple—the intensity is just the sum of the intensities of all the pixels defined as part of the spot. However, there are several refinements that are usually done before this sum is calculated: (1) A background value is subtracted before the intensities are summed. Ideally, after a fluorophore has bleached, the summed intensity should drop to approximately zero, if the background was effectively subtracted. (2) A *weighted* sum is used by some groups (including the Ha lab). If the IntensityWeightGauss option is 1 in TracesSetup, which I highly recommend, each spot's intensity will be weighted, in each frame, with a 2D gaussian centered on the spot and having the amplitude you set in GaussWeightAmp. This has the effect of smoothing out some of the fluctuations in a spot's intensity due to noise. The Ha lab uses an amplitude of 2, which just stretches the results out along the intensity axis. Also, you can either weight with a Gaussian whose width is fixed to the values set in FixSpotVar; or you can weight each spot by a Gaussian whose width was calculated when the spot was originally found. The Ha lab fixes the spot variances; I've found it doesn't make a huge difference. Using spot-specific variances can be nice if part of the field of view has distortions that result in more ellipsoid, rather than round, spots. (Note that if you want ellipsoid spots, the UseSymGauss option must be 0.)

²CalcBgknd actually uses the built-in Matlab function imopen, which is much faster than the options for doing a median filter and results in almost the same thing.

5 Extras: HMM analyses of Traces outputs.

Hidden Markov model (HMM)-based approaches are some of the most popular methods for extracting quantitative information from FRET-vs-time trajectories. The “extras” subdirectory in Traces includes a function called “ConvertTovbFRETInput.m,” which converts the output files saved in the Traces GUI to the input format for the commonly-used vbFRET hidden Markov modeling software [2]. We have found vbFRET to be most useful for small data sets (it is relatively slow), and for data that includes fast switching between two or three states.

For our purposes, though, we have instead adapted a free, open-source HMM python library called pyhsmm for use with the output files of Traces. *Note: our adaptation of pyhsmm is not as well-documented or as thoroughly tested as Traces!* Technically, pyhsmm is the name of the python library. However, in the following we will refer to the code we wrote to use this library to analyze the output of Traces as “pyhsmm” also. Our code can be found in the “pyhsmm_for_Traces” subdirectory in “extras”.

5.1 pyhsmm with Traces: First-time setup

Additional information on running pyhsmm on the output of Traces can also be found in the README that accompanies the sample data at stephlj/TracesSampleData repository.

5.1.1 Installing pyhsmm.

pyhsmm can be downloaded from <https://github.com/mattjj/pyhsmm>. Installation instructions can be found in the README in the pyhsmm repo. You will also need a python installation.

5.1.2 Configuring pyhsmm to load the output of Traces.

(1) Tell pyhsmm where to look for data and where to save the results:

Open config.py and edit *datadir* to point to the master directory for your Traces output files; edit *resultsdir* to point to where you want pyhsmm to store the HMM results.

As an example, let’s say I want to analyze all of my 51 nM SNF2h with 1 mM ATP data. Say I collected data on two separate days and saved the analysis of that data in

```
/Users/Steph/Documents/smFRETdata/TracesOutput/SNF2h51nMATP1mM_2015_May05  
/Users/Steph/Documents/smFRETdata/TracesOutput/SNF2h51nMATP1mM_2015_Jun01
```

Each of these subdirectories can contain the analysis of multiple acquisitions from Micro-Manager. Let’s suppose that on May 5, I took two movies; then SNF2h51nMATp1mM_2015_May05 will have Spot1_1.mat, Spot1_25.mat, ... Spot2_100.mat, Spot2_250.mat, etc.

In config.py, datadir would be

```
/Users/Steph/Documents/smFRETdata/TracesOutput  
resultsdir might be  
/Users/Steph/Documents/smFRETdata/pyhsmmOutput
```

(2) Create a subdirectory in *resultsdir* for data that you want to analyze together. The name of this directory will be the input to ExamineHMMResults() or the second argument if you run auto_analysis from the command line; it will be called “out_dir” in the rest of this manual.

In the above example, out_dir could be “SNF2h51nMATP1mM.” The full path to out_dir would be

```
/Users/Steph/Documents/smFRETdata/pyhsmmOutput/SNF2h51nMATP1mM
```

(3) In *out_dir*, create a text file called “goodtraces.txt.” Each line of goodtraces should list a .mat file saved by Traces that represents a single trajectory (*Spot*.mat), along with its parent directory. See the example goodtraces.txt file included with this repo for more details.

*goodtraces.txt tells pyhsmm what data to load, and any information about cropping or excluding particular traces. NOTE: Once you run ExamineHMMResults or auto_analysis, pyhsmm will generate a set of output files detailed below, including a *_Results.mat file for each line in goodtraces.txt that does not start with a #. If you then add a # to a line in goodtraces.txt, to remove the corresponding file from analysis, pyhsmm will not operate on that file, but the old output files will need to be removed manually!*

(4a) If you are going to call auto_analysis from the command line, create a directory in *out_dir* called “ResultsFigs”.

(4b) If you are going to use ExamineHMMResults() in Matlab to run the analysis, open ExamineHMMResults and edit the parameters at the top of the function.

Note: to my knowledge, using ExamineHMMResults to call auto_analysis by “shelling out” of Matlab’s IDE only works on a Mac, at least as currently implemented.

5.1.3 Configuring models and model parameters.

pyhsmm is a general python library for fitting HMMs and related models to many kinds of noisy data, not just single-molecule FRET data. As such, there are many ways to customize your use of it to suit your particular needs.

In our case, we use pyhsmm to fit HMMs with Gaussian emissions and instantaneous transitions between states; so in fret.py you will notice that we import Gaussian from pyhsmm.basic.distribution. We use a hierarchical Dirichlet process (HDP) prior to automatically select the number of states in each trajectory; and we also make use of the “sticky” regularizer to penalize very short-lived states (which in our case are usually noise states that we want the model to ignore). Thus we import WeakLimitStickyHDPHMM from pyhsmm.models. Obviously, if your data have very fast switching events between states, or the noise is non-Gaussian, you will want to import a different built-in distribution from the pyhsmm library, or exclude or change the sticky regularizer, etc.

Regardless, in fret.py you may want to change the hyperparameters for the Gaussian emissions (see line 70, “obs_hypparams = ...” in make_model); and/or the value of the sticky regularizer, line 84, where the model is defined (also in make_model). This is also where you would change the kind of model if you want to use something other than WeakLimitStickyHDPHMM. See the comments in fret.py for more information.

Lastly, if you want to change the way data are loaded, the figures that are generated (see below), what gets saved to the .mat files that are generated (also see below), etc, you would do that in fret.py.

5.2 Running pyhsmm on Traces output files.

5.2.1 Running pyhsmm from the command line.

Navigate to the folder containing auto_analysis.py, FRET.py and config.py. In a command line (*e.g.*, in a Terminal on a Mac), run

```
python auto_analysis.py out_dir  
where out_dir is, continuing the example from above, SNF2h51nMATP1mM.
```

5.2.2 Running pyhsmm from Matlab with the GUI provided in Traces/extras.

In Matlab, call `ExamineHMMresults('out_dir')`. If everything is set up correctly, a series of “Done with <each line from goodtraces.txt>!” will be displayed. Once all the files are analyzed, a rudimentary GUI like the one included with Traces will appear, allowing the user to page through the results of each file’s analysis. Through the GUI, you can discard or un-discard a trace, zoom and unzoom, and re-run the pyhsmm analysis for this particular trace.

The GUI will display four plots: the top two show the red and green acceptor intensities, and the bottom to the calculated FRET values. The left plots show the data stored in “RedI,” “GrI”, and “FRET” in the Traces output file (so including any smoothing for visualization that the user implemented for viewing in the Traces GUI); the right plots show the data in “unsmoothedRedI” *etc.* Note that the HMM analysis is performed on the unsmoothed data; sometimes the HMM segmentation will look wrong, until you examine the unsmoothed data that it actually works with.

To change the start and end cropping, you will have to edit the `goodtraces.txt` file directly. You can do this while the Matlab GUI is running; if you edit `goodtraces.txt`, use the “r” option in the GUI to re-run the analysis. The `goodtraces.txt` file will be re-loaded by `auto_analysis`, and the new cropping information used in the resulting output.

5.3 pyhsmm outputs.

`auto_analysis` will generate a set of .png files in the `ResultsFigs` subdirectory, which show the unsmoothed input data with the HMM segmentation overlaid on it. It will also generate a .mat file for every line of `goodtraces.txt` that does not start with #, which has all of the elements of the Traces output .mat files, plus any start or end crop values indicated in `goodtraces.txt`, and:

model_data: Frames-by-2 matrix, where the first column contains the donor dye intensities used as the input to `pyhsmm`, and the second column are the acceptor intensities;

model_durations: Durations, in frames, of each state fit by the HMM. `sum(model_durations) = size(model_data,1)`.

model_redgreenseq and *model_fretseq*: Frames-by-2 and 1-by-frames matrices, respectively, containing for each frame the *fitted* donor (“green”), acceptor (“red”), and FRET values for that frame. As an example: imagine a trajectory starts in high FRET, $\text{FRET}=0.95$, and the fitted HMM has this FRET state continue for 47 frames. Then `model_durations(1) = 47` and the first 47 elements of `model_fretseq` will have identical values of 47.

discard: If this is 0, the user indicated this trace should be discarded from further analysis. If absent or 1, keep this trace. *Note: this field is only generated through ExamineHMMresults.*

model_stateseq_norep: This vector tells you the order of states assigned by the HMM (and how many there were). For example, if `model_stateseq_norep = [7, 0, 7, 4, 7, 0]`, the fitted HMM has three unique states, which the algorithm called state 7, state 0 and state 4. The first `model_durations(1)` frames in the trajectory were assigned to state 7, which has FRET value `model_fretseq(1)`; the next `model_durations(2)` frames were assigned to state 0, which FRET value `model_fretseq(model_durations(1)+1)`; the next `model_durations(3)` frames were assigned to the initial state, state 7; and so on.

model_stateseq: Same as `model_stateseq_norep` but has one state assignment per frame.

In addition, `fret.py` also generates a `modelfits.pkl` file that contains the output of `pyhsmm`, if you don’t want to use .mat files.

A note about the outputs of pyhsmm: Many of these outputs are type int32, NOT double! Be careful, especially if calling auto_analysis through the command line, that you convert them to double precision before manipulating them in a way that would generate non-integer values (for example, dividing by fps to get seconds instead of frames).

Lastly, if you call pyhsmm through Matlab by running ExamineHMMResults, a .mat file called ToAnalyzeFurther.mat will be generated. This file contains a single vector, called tokeep, that is the same length as the number of *_Results.mat files, and has a value of 1 if the corresponding trajectory is to be kept for further analysis, but 0 if it was selected to be discarded using the “d” option in the ExamineHMMResults GUI. This is a very fragile way of keeping track of which trajectories were flagged as to be discarded in the ExamineHMMResults GUI! If you add or remove *_Results.mat files, ExamineHMMResults will warn you that the current number of results files does not match the length of tokeep, and if you continue, will overwrite tokeep, setting all values to 1 for the new round of analysis. It is therefore more robust to discard traces by adding a # to the start of the line in goodtraces.txt.

6 Details and derivations.

6.1 Channel mapping/transformations: General concepts

The essence of finding a map that relates locations in the donor channel to locations in the acceptor channel, and vice versa, is finding a transformation in the mathematical sense. We will be particularly concerned with massaging the desired channel maps into *linear* transformations, because linear transformations can be described by matrices, and we can bring the power of linear algebra to bear on them. In particular, for a set of points (think spots—beads or dyes—in one of the channels) $\vec{x} = \{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_n\}$, where the Cartesian coordinates of each point are given by $\vec{x}_i = (x_{i1}, x_{i2})$, we can write a linear transformation to the other channel in terms of a matrix \mathbf{A} , such that

$$\vec{y} = \mathbf{A}\vec{x}, \quad (1)$$

where \vec{y} are the Cartesian coordinates of the points in the other channel. We will use the same convention for \vec{x} and \vec{y} as in the code, namely, that a single spot's coordinates will be a column vector, and so a set of spots will be a $2 \times n$ matrix, where n is the number of spots. Then we know that in the simple example of Eqn. (1) above, \mathbf{A} will be a 2×2 matrix, though this will not be the case for all of the transformations we describe below.

The goal, of course, is to find \mathbf{A} , which will be the map describing how to find a spot's location in the other channel, if we know where it is in the first channel. More precisely, what we want is a *best fit* value for \mathbf{A} . That is, we want to find the \mathbf{A} that minimizes the error of our estimate of \vec{y} , given \vec{x} ; i.e. we want to find \mathbf{A}^* such that

$$\mathbf{A}^* = \underset{\mathbf{A}}{\operatorname{argmin}} \|\vec{y} - \mathbf{A}\vec{x}\|_2^2. \quad (2)$$

According to Eqn. (2), \mathbf{A}^* will be the \mathbf{A} that minimizes the difference between the true positions of the points represented by \vec{y} , and where we estimate them to be based on applying matrix \mathbf{A} to \vec{x} . (The notation $\|\dots\|_2$ just means the “norm-2” (Euclidean norm or Euclidean distance) of whatever is between the $\|\cdot\|$'s; in 2-space, you will recall that for two points (x_1, x_2) and (y_1, y_2) , their Euclidean distance is $\sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2}$.)

How do we find \mathbf{A}^* ? We use the **normal equations** as conditions of optimality. I refer you to a linear algebra textbook that discusses the normal equations for more details, but, in short, using the normal equations as conditions for optimality means we want to find \mathbf{A}^* such that

$$(\vec{y} - \mathbf{A}^*\vec{x}) \cdot \vec{x}^T = \vec{0}. \quad (3)$$

(If you want some intuition about what's going on here, think about reading Eqn. 3 row-wise: The first row is an orthogonality statement that constrains the first row of \mathbf{A} , saying that the first row of the residual, that is, the stuff in the parenthesis on the left-hand side, should be orthogonal to the row space of \vec{x} . The second row of Eqn. 3 says that the second row of the residual should *also* be orthogonal to the row space of \vec{x} —meaning we have two orthogonality constraints that must be satisfied on \mathbf{A}^* .)

Now what we need to do is solve Eqn. (3) for \mathbf{A}^* . Rearranging Eqn. (3) and moving the term containing \mathbf{A}^* to the other side, we have

$$\vec{y}\vec{x}^T = \mathbf{A}^*\vec{x}\vec{x}^T. \quad (4)$$

Remember these are matrices, so we cannot “divide” both sides by $\vec{x}\vec{x}^T$, but we can multiply both sides of the equation, on the righthand side, by the inverse of $\vec{x}\vec{x}^T$. Is $\vec{x}\vec{x}^T$ invertible? You can show

that it is (which I'm not going to do here, but you can Wikipedia linear least squares and/or look up the normal equations in a linear algebra textbook if you'd like to know more). So we have our final solution that

$$\vec{y}\vec{x}^T(\vec{x}\vec{x}^T)^{-1} = \mathbf{A}^*. \quad (5)$$

In summary, to find the best-fit transformation that takes a set of spots \vec{x} in one channel to a set of spots \vec{y} in another channel, we calculate the lefthand side of Eqn. 5. In practice, Matlab has a smarter way of solving these equations, represented by the slash operator. So in the channel mapping code of this repository, you will find \mathbf{A}^* computed as $\vec{y}\vec{x}^T/(\vec{x}\vec{x}^T)$ rather than as written in Eqn. 5.

6.2 Affine versus polynomial maps.

Before we describe how to find \mathbf{A} , or more precisely, how to fit it using a set of control points whose locations we know in both channels, we first give some intuition about what the different transformations implemented in this code do to a set of spots.

6.2.1 Affine transformations.

Affine transformations allow for stretching/shrinking, rotation, shear, translation, and reflection (which isn't really applicable to TIRF microscopy). An affine transformation can be represented as

$$\vec{y} = \mathbf{M}\vec{x} + \vec{b} \quad (6)$$

Fig. 9 shows how the elements of \mathbf{M} and \vec{b} accomplish the various changes possible with affine maps. As we will see below, an affine transformation describes donor-acceptor pairs of spots in an smFRET experiment fairly well, though a polynomial transformation tends to do slightly better (in our hands as well as in others'—see, for example, [3, 1]).³

6.2.2 Polynomial transformations.

Polynomial transformations represent a more general and less well-defined class of transformations than affine transformations, and in fact, as we will see, even the terms used to describe them in the IDL versus Matlab documentation differ in significant ways.

A polynomial transformation is one in which each element of the output vector, y_{j1} and y_{j2} , is determined by a polynomial of the input vector elements, x_{i1} and x_{i2} . The polynomials for y_{j1} and y_{j2} can be different. An example of a very simple polynomial transformation is

$$\begin{aligned} y_1 &= 3 + x_1 + 2x_2 \\ y_2 &= 1 + 5x_1 + 3x_2 \end{aligned} \quad (7)$$

Each term (3 , x_1 , $2x_2$, etc) is called a **monomial**. Note that this is actually a linear system of equations, and therefore is in fact an affine transformation—it can be written in matrix form as in Eqn. (6) above. This means that polynomial transformations can do to a set of points all of the things shown in Fig. 9, as well as other things.

To be a polynomial transformation that is not affine, the **degree** of one or both of the polynomials that define the transformation must be greater than 1. Degree is defined in two ways: the **maximum**

³We note here that the Ha lab IDL code first generates an affine map, by asking the user to pick three true pairs of spots (three being the minimum number of points needed to define an affine transformation), and uses this to automatically match more spots with which to generate a polynomial map.

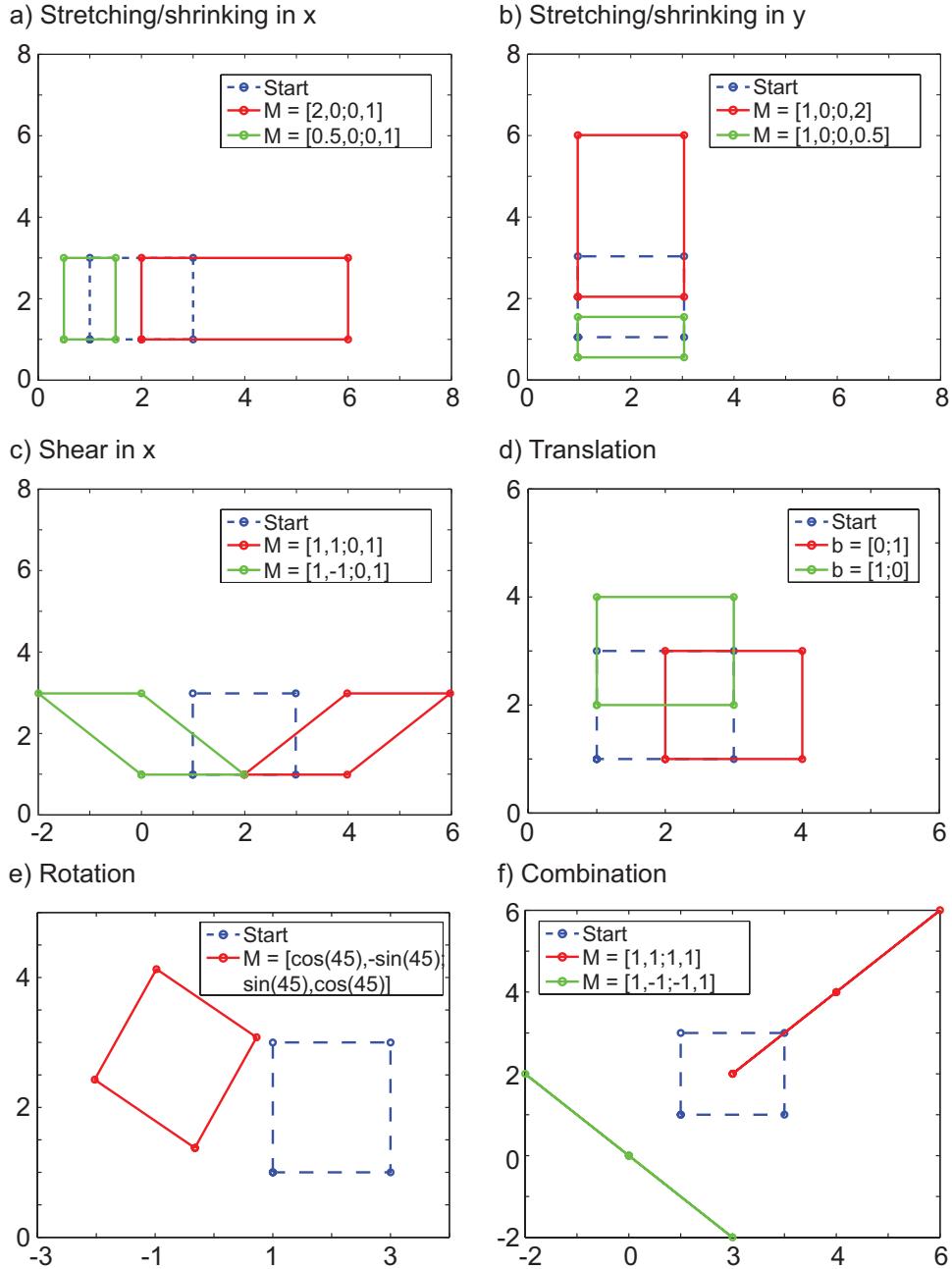


Figure 9: **Examples of affine transformations.** In all panels, the blue box outlined in dashed lines represents \vec{x} , the starting set of points. The legends show the choice of \mathbf{M} or \vec{b} that gives the transformed set of points \vec{y} , shown in green or red. Where not listed, $\mathbf{M} = [1, 0; 0, 1]$; and $\vec{b} = [0,0]$.

degree is the highest power of a single variable that appears in the polynomial; for example, in the polynomial $3 + x_1x_2 + x_1^2x_2$, the maximum degree is 2. The **total degree** is the highest power of combinations of variables that appears; for example, in the polynomial $3 + x_1x_2 + x_1^2x_2$, the total degree is 3.

This distinction becomes important when we look at the polynomial transformations that can be fit to sets of control points using IDL's POLYWARP function versus Matlab's fitgeotrans function. IDL defines the degree of the polynomial to be fit as the *maximum degree*, whereas Matlab defines the degree of the polynomial to be fit as the *total degree*. The Ha lab IDL code uses POLYWARP to fit a “third-degree” polynomial to control points to find the channel mapping; Matlab’s fitgeotrans only implements total degree 2, 3, or 4 polynomials, which means we cannot use Matlab’s built-in functionality to fit the same kind of polynomial as in the Ha lab code. To be specific, POLYWARP with a degree 3 polynomial allows terms like $x_1^3x_1^2$, because the maximum degree for that polynomial is 3; but because its total degree is 5, this monomial would not be included in a degree-4 polynomial fit (or a degree-3 fit) by Matlab’s fitgeotrans.

We first specifically define polynomial transformations according to IDL’s POLYWARP and Matlab’s fitgeotrans. POLYWARP finds parameter vectors \vec{K}_x , \vec{K}_y such that, for a starting set of points $[x_1 \atop x_2]$ and an ending set of points $[y_1 \atop y_2]$,

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} \sum_{i,j=0}^D K_{x_{i,j}} x_1^i x_2^j \\ \sum_{i,j=0}^D K_{y_{i,j}} x_1^i x_2^j \end{bmatrix}, \quad (8)$$

where D is the degree of the polynomial fit (set by the user). Note that for a maximum degree of 3, the highest-order term possible here is total degree 6, namely $x_1^3x_2^3$. More explicitly, we can write out Eqn. (8) as

$$y_1 = K_{x_{00}} + K_{x_{10}}x_1 + K_{x_{01}}x_2 + K_{x_{11}}x_1x_2 + K_{x_{20}}x_1^2 + \dots \quad (9)$$

$$y_2 = K_{y_{00}} + K_{y_{10}}x_1 + K_{y_{01}}x_2 + K_{y_{11}}x_1x_2 + K_{y_{20}}x_1^2 + \dots \quad (10)$$

Although POLYWARP returns \vec{K}_x , \vec{K}_y as vectors, we can think of them as matrices, with the rows corresponding to the coefficients for each monomial in x_1 (x^0 (=constant), x , x^2 , ...) and the columns corresponding to the coefficients of each monomial in x_2 . That is, for a degree 3 polynomial fit, we can arrange the elements of \vec{K}_x as

$$\begin{array}{ccccc} & \mathbf{x}_2^0 & \mathbf{x}_2^1 & \mathbf{x}_2^2 & \mathbf{x}_2^3 \\ \mathbf{x}_1^0 & K_{x_{00}} & K_{x_{01}} & K_{x_{02}} & K_{x_{03}} \\ \mathbf{x}_1^1 & K_{x_{10}} & K_{x_{11}} & K_{x_{12}} & K_{x_{13}} \\ \mathbf{x}_1^2 & K_{x_{20}} & K_{x_{21}} & K_{x_{22}} & K_{x_{23}} \\ \mathbf{x}_1^3 & K_{x_{30}} & K_{x_{31}} & K_{x_{32}} & K_{x_{33}} \end{array} \quad (11)$$

and similarly for \vec{K}_y . Note that there will therefore be 32 coefficients (16 for K_x , 16 for K_y) for a degree 3 polynomial fit in IDL.

On the other hand, Matlab’s documentation (see `images.geotrans.PolynomialTransformation2D` class) tells us that a degree 2 polynomial fit returns a \vec{K}_x vector of length 6 for a 2-degree fit, length 10 for degree 3, and length 15 for degree 4. This means that a third degree fit in Matlab will have only 20 parameters, rather than the 32 of IDL. Even the degree 4 fit will have only 30 parameters; though this will result in a fit of a comparable number of parameters to the IDL degree-3 fit, the particular monomials that are allowed in that fit will be different. In particular, a third-degree fit in Matlab will

return a \vec{K}_x vector that would correspond to the matrix

$$\begin{array}{ccccc} & \mathbf{x}_2^0 & \mathbf{x}_2^1 & \mathbf{x}_2^2 & \mathbf{x}_2^3 \\ \mathbf{x}_1^0 & K_{x_{00}} & K_{x_{01}} & K_{x_{02}} & K_{x_{03}} \\ \mathbf{x}_1^1 & K_{x_{10}} & K_{x_{11}} & K_{x_{12}} & - \\ \mathbf{x}_1^2 & K_{x_{20}} & K_{x_{21}} & - & - \\ \mathbf{x}_1^3 & K_{x_{30}} & - & - & - \end{array} \quad (12)$$

where we have indicated monomials that are not included in the fit by a dash. Note that whichever of these polynomials you fit, you will need more than 3 pairs of spots to fit them—for example, you will need at least 16 spots to fit a polynomial that needs 32 coefficients!

6.3 Finding non-linear transformations.

In Sec. 6.1 we described how to find the best-fit *linear* transformation, given a set of control points whose positions we knew with reasonable accuracy in both channels. Comparing Eqns. (3) and (6), it is not immediately obvious how to use the normal equations to find an affine transformation, and it is even less clear how to find a polynomial transformation.

What we need to do is to embed the affine and polynomial transformations in a linear space, such that they take the form of Eqn. (1). It is actually relatively straightforward to do this for an affine transformation. Redefine \vec{x} , \vec{y} , \mathbf{M} , and \vec{b} as \vec{x}' , \vec{y}' , and \mathbf{A} such that

$$\vec{x}' = \begin{bmatrix} \vec{x} \\ 1 \end{bmatrix}, \quad (13)$$

$$\vec{y}' = \begin{bmatrix} \vec{y} \\ 1 \end{bmatrix}, \quad (14)$$

and

$$\mathbf{A} = \begin{bmatrix} \mathbf{M} & \vec{b} \\ 0 & 1 \end{bmatrix}. \quad (15)$$

(All we're doing is adding a row of one's to \vec{x} and \vec{y} , and combining \mathbf{M} and \vec{b} into a new matrix along with a new row of two 0's and one 1. Note that for an affine transformation in 2-space, \mathbf{M} is 2x2 and \vec{b} is 2x1—see also Fig. 9.) It should be clear that now $\vec{y}' = \mathbf{A}\vec{x}'$ is a linear transformation, whose solution for a best-fit \mathbf{A} is given in Eqn. (5), and that it is equivalent to the affine transformation of Eqn. (6).⁴ (If that second point is not immediately clear, do the matrix multiplication for $\mathbf{A}\vec{x}'$ and see that the first row of the resulting matrix is $\mathbf{M}\vec{x} + \vec{b}$.)

We can do a similar thing to embed a polynomial transformation in a linear space, with some slight modifications. Instead of the redefinitions above, we will instead define a new variable $\vec{\tilde{x}}$ such that

$$\vec{\tilde{x}} = \begin{bmatrix} 1 \\ x_1 \\ x_2 \\ x_1 x_2 \\ x_1^2 \\ \dots \end{bmatrix}, \quad (16)$$

⁴Note that Matlab's built-in affine transformation calculation function (cp2tform or fitgeotrans, depending on your version of Matlab) uses a different convention than I use here. Matlab computes the transformation: $\text{transpose}(\vec{y}') = \text{transpose}(\vec{x}')\mathbf{A}_{\text{Matlab}}$, such that $\mathbf{A} = \text{transpose}(\mathbf{A}_{\text{Matlab}}^{-1})$, where \mathbf{A} is the transformation matrix in my notation.

up to/including as many monomials as we want to include in our transformation, and where as a reminder (x_1, x_2) are the Cartesian coordinates of each point in the starting vector \vec{x} . Note then that if our transformation included, say, 20 monomials, and we have n points in \vec{x} , $\vec{\tilde{x}}$ would be a $20 \times n$ matrix. We define a new variable \vec{y} similarly. Then our transformation matrix $\tilde{\mathbf{A}}$ will contain the coefficients of the polynomials that relate the elements of the input vector (x_1, x_2) to those of the output vector (y_1, y_2) .

Take, for example, the polynomial transformation

$$\begin{aligned} y_1 &= 3 + x_1 + 2x_1x_2 + 4x_2^2 \\ y_2 &= 1 + 5x_1x_2 + 3x_1x_2^2 \end{aligned} \quad (17)$$

We would then have

$$\vec{\tilde{x}} = \begin{bmatrix} 1 \\ x_1 \\ x_1x_2 \\ x_2^2 \\ x_1x_2^2 \end{bmatrix} \quad (18)$$

and

$$\tilde{\mathbf{A}} = \begin{bmatrix} 3 & 1 & 2 & 4 & 0 \\ 1 & 0 & 5 & 0 & 3 \end{bmatrix}. \quad (19)$$

As with the embedding of the affine transformation in a linear space above, it should be clear from Eqns. (18) and (19) that $\vec{y} = \tilde{\mathbf{A}}\vec{\tilde{x}}$ is linear, and that $\tilde{\mathbf{A}}\vec{\tilde{x}}$ represents the polynomial transformation given by Eqn. (17) above.

Finally we note that in the case of polynomial transformations, the transformation must be explicitly calculated in both directions (acceptor to donor and donor to acceptor), whereas affine transformations are invertible and the transformation in both directions can be calculated, if the transformation is known for one direction.

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