How to use the python coulter counter fitting script

Updated 21/08/2022

Based on initial script from Amy Briffa, however it has since been significantly modified by Rose McNelly

The script will be published in an upcoming publication, McNelly et al. 2023.

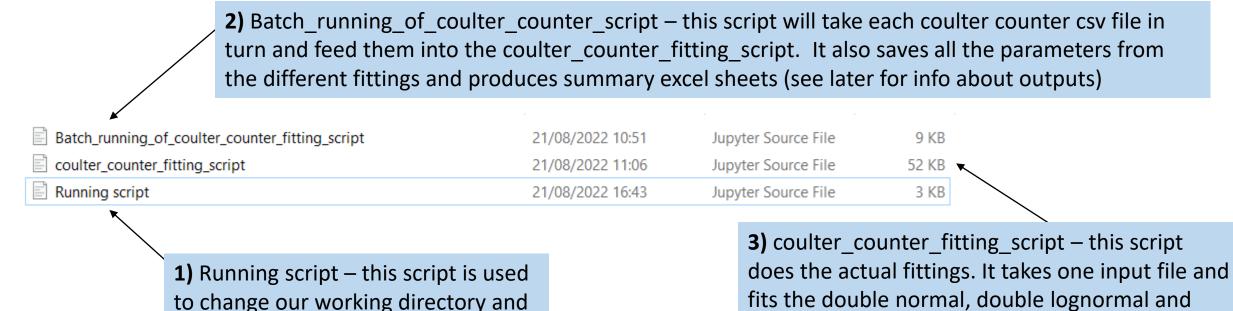
What does the script do?

- The script fits 3 different models to bimodal coulter counter traces:
- 1) Two normal curves (double normal)
- 2) Two lognormal curves (double lognormal)
- 3) Lognormal for B granule peak and normal for A granule peak (lognormal-normal)

• It compares the different fittings using parameter uncertainties and a standard error of regression test. It produces estimates of A and B granule diameter and B granule content (as a ratio) for each model.

What are the scripts?

submit the other two scripts



Why can't we just use the coulter_counter_fitting_script by itself? Technically we could but in order to get excel sheets with all the parameters in we need to use script 2 (batch_running_of_coulter_counter_script). It also means that if we have multiple coulter counter files to analyse at once we only need to submit the script once and not for every input file. The running script also makes it easier so that we can tell the computer where are input files are (so that we don't have to copy them into our Jupyter lab directory – see later for info about Jupyter lab).

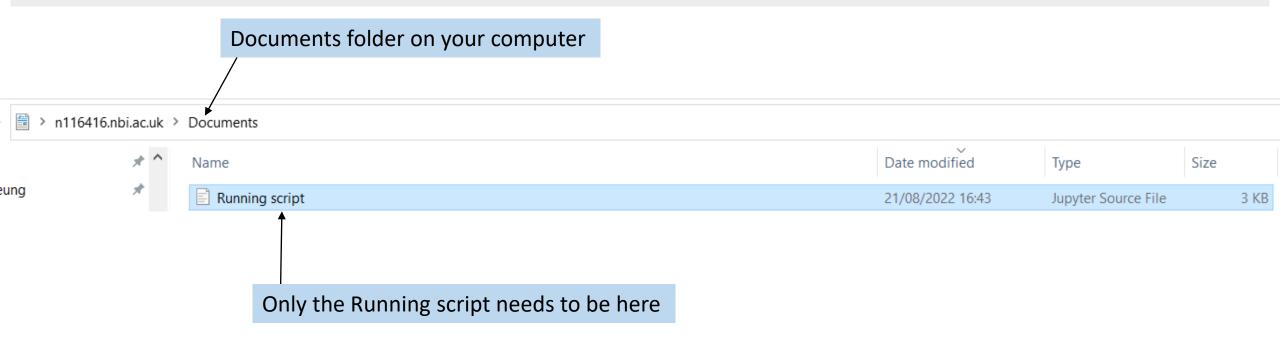
lognormal-normal curves to the data. It produces

a PDF with graphs of the 3 different fittings and a

comparison of the 3.

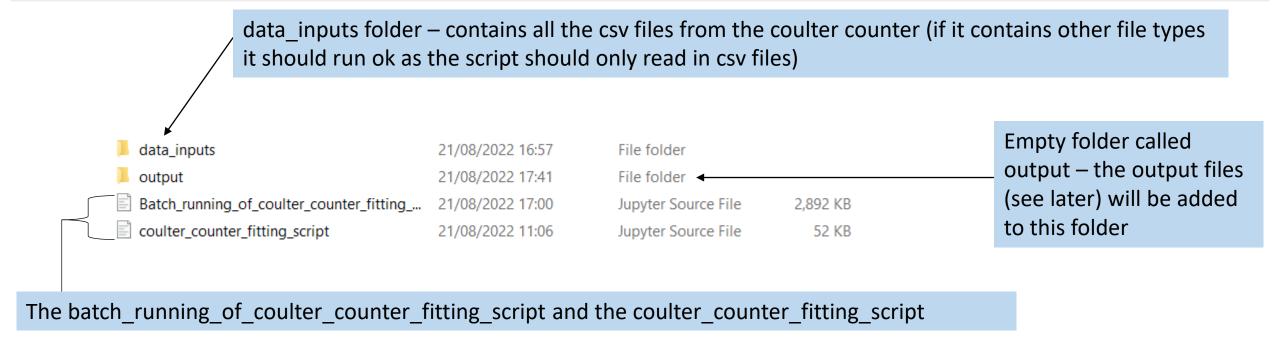
How should my data be organised?

I always place the Running script (only this has to be here) in my Documents folder as this is easy to navigate to using Jupyter notebook (see later)



How should my data be organised?

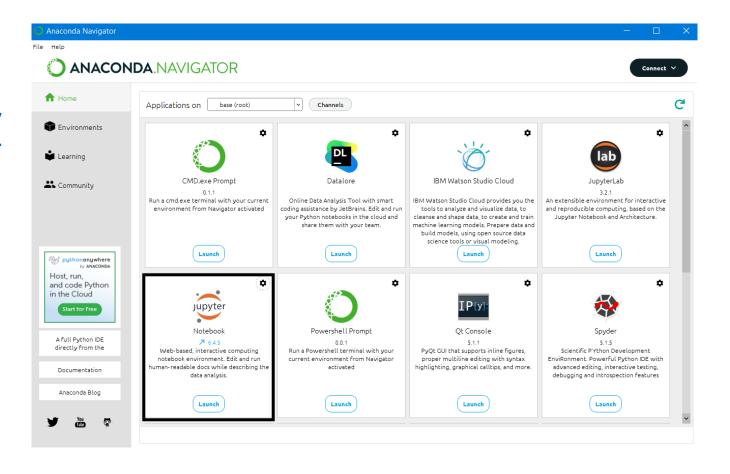
Everything else can be in a different folder, this can be anywhere and doesn't have to be within your Documents folder, but if you want it in your Documents folder this is fine too. It should be organised as shown here:



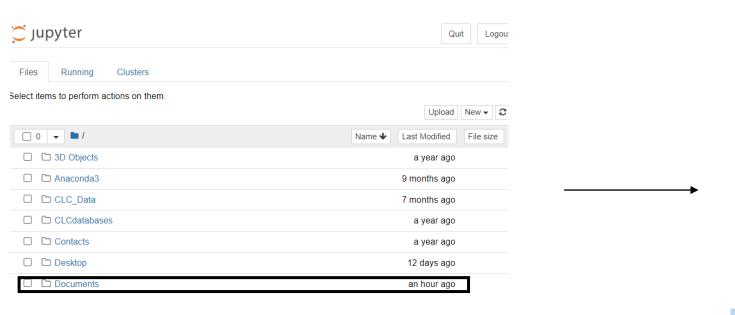
The folder which is organised like this is you working directory, take note of the name of this folder and where it is (the full path) as you'll need it

1) Download Anaconda Navigator

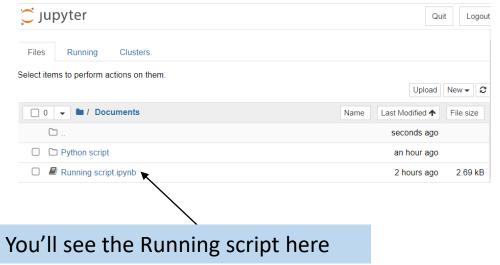
(https://docs.anaconda.com/ anaconda/install/) and open Jupyter Notebook (this will open in an internet browser)

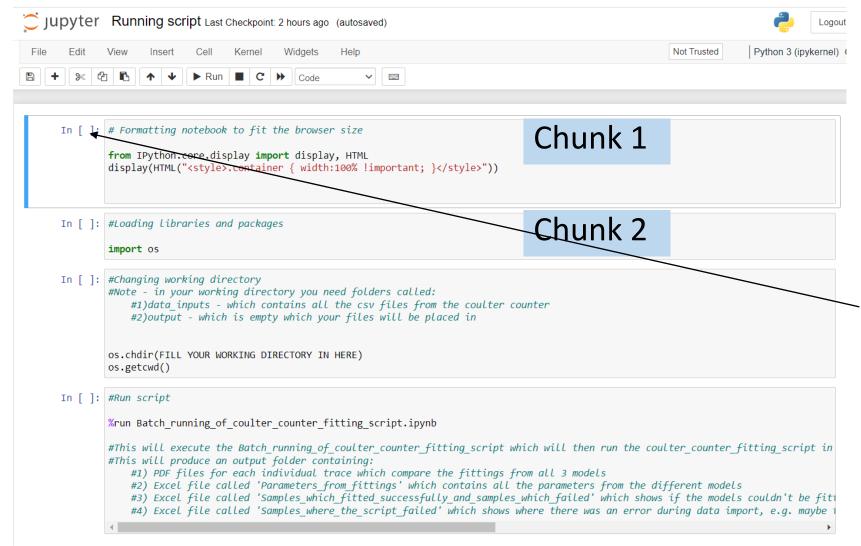


2) Navigate to your Documents folder



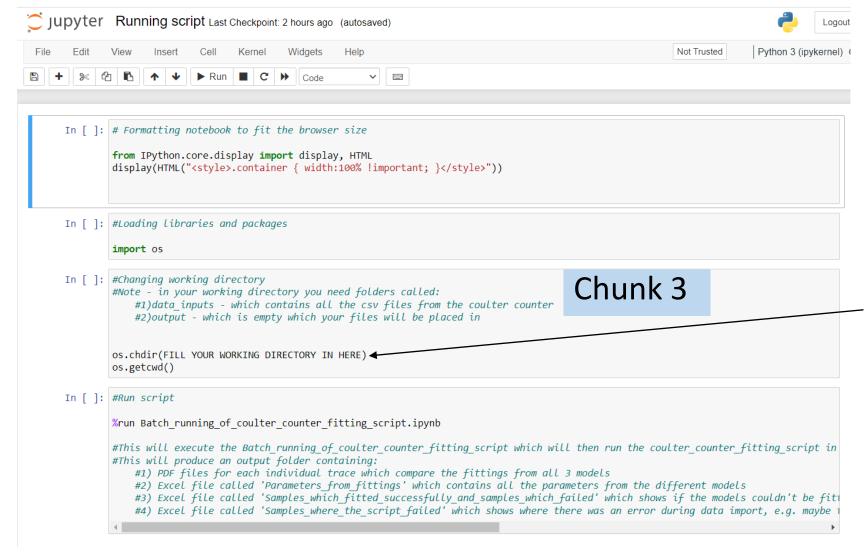
3) Open the Running script





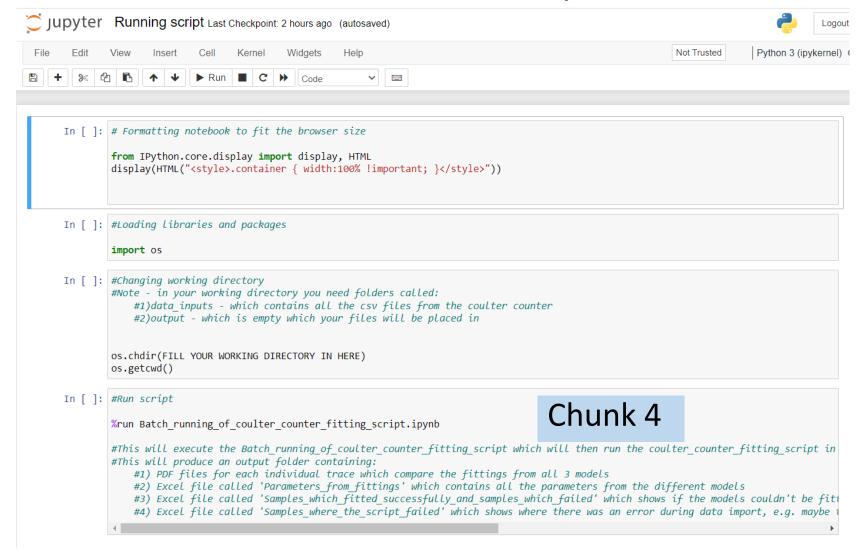
4) Run the first two chunks (shown as grey boxes), by clicking in each chunk and pressing Ctrl and Enter.

If it has run successfully a small number should appear in the square brackets on the left of the chunk



5) In chunk 3, change your working directory to the location which contains the other scripts, your data_inputs and output folder. You need to put in the full path, not just the folder name, and put it within quotation marks, e.g. 'U:\Working_Directory_Folder'

Run the chunk, if it has worked then it should print the working directory folder path directly underneath the chunk.



6) Run chunk 4 – this submits the Batch_running_of_coulter _counter_script which itself will submit the coulter_counter_fitting_s cript (you won't have to do anything)

You'll start to see the outputs appear in the output folder

If you want to know how the script works see later

What are the outputs?

Parameters_from_fittings – contains all the parameters from the 3 different fittings and also contains the uncertainties and the standard error of regression

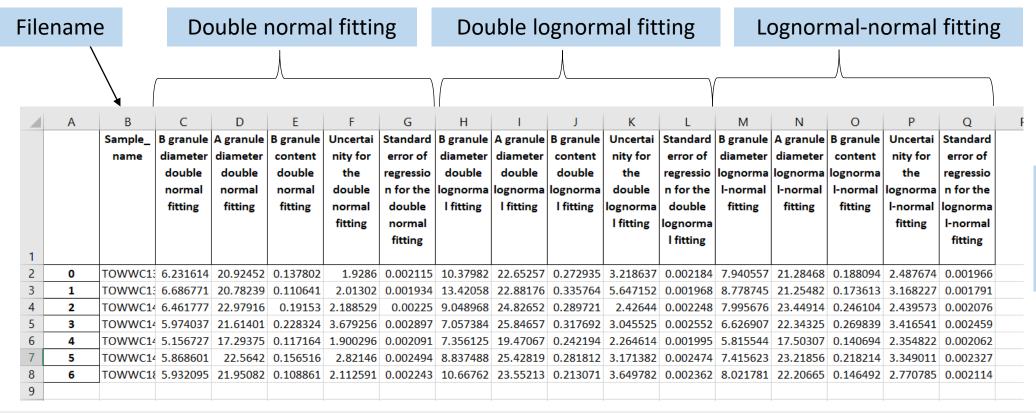
Samples_where_the_script_failed – contains a list of filenames where the script failed (most likely during data import), if a filename appears in this spreadsheet it could be because the coulter counter CSV file is corrupted or not in the correct format

because the coulter counter CSV file is	corrupted or not in the	e correct format	
Parameters_from_fittings	21/08/2022 16:58	Microsoft Excel W	8 KB
Samples_where_the_script_failed	21/08/2022 16:58	Microsoft Excel W	6 KB
Samples_which_fitted_successfully_and_samples_which_failed	21/08/2022 16:58	Microsoft Excel W	6 KB
★ TOWWC138_rep1_26 Jan 2021_104.#m4	21/08/2022 16:58	Adobe Acrobat D	49 KB
★ TOWWC139_rep1_26 Jan 2021_107.#m4	21/08/2022 16:58	Adobe Acrobat D	49 KB
★ TOWWC140_rep1_26 Jan 2021_111.#m4	21/08/2022 16:58	Adobe Acrobat D	48 KB
★ TOWWC141_rep1_26 Jan 2021_112.#m4	21/08/2022 16:58	Adobe Acrobat D	49 KB
★ TOWWC142_rep1_26 Jan 2021_113.#m4	21/08/2022 16:58	Adobe Acrobat D	48 KB
★ TOWWC143_rep1_26 Jan 2021_114.#m4	21/08/2022 16:58	Adobe Acrobat D	49 KB
★ TOWWC183_rep1_19 Jan 2021_02.#m4	21/08/2022 16:58	Adobe Acrobat D	49 KB

Samples_which_fitted successful and sam ples which failed contains a list of filenames for each of the different fitting methods for samples which worked and samples where the fitting failed. The fitting might have failed if the data is not bimodal in shape.

PDF files – contain graphs of all the different fittings

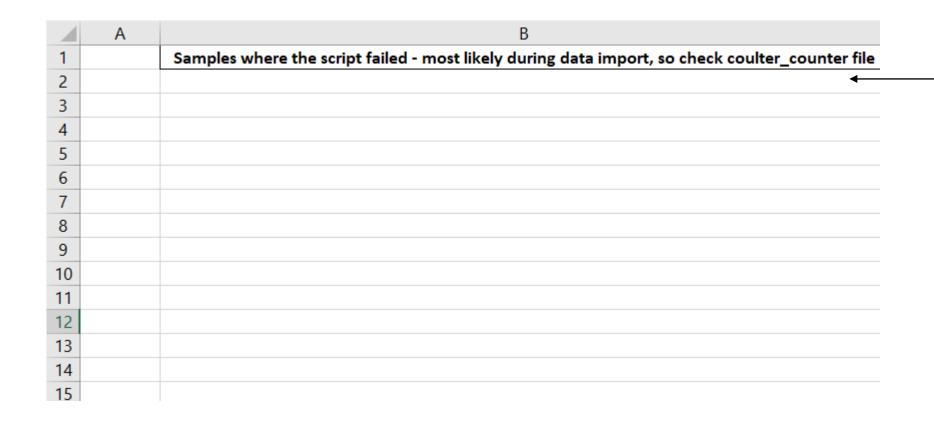
Parameters_from_fittings



Each row contains the parameters from a different input file

Uncertainty is a measure of how sure we can be about the parameters. Higher uncertainties mean that the parameters in the models could take lots of different values and still result in the same fitting. Therefore <u>we want the uncertainty to be as low</u> <u>as possible</u> as this means we are more certain that the model parameters are 'correct'. Standard error of regression is a form of R² for non-linear models. It measures how well the predicted model fits the data. Higher standard errors of regression means that the data deviates from the model more. So <u>we want the standard error of regression to be as low as possible</u>.

Samples_where_the_script_failed



If any of the input files failed to load into the script properly they will be listed in this column.

Here this column is empty so all the files were inputted correctly.

NOTE – this does NOT mean that the fitting won't work as it has failed before the fitting has been attempted.

Samples_which_fitted_successful_and_samples which failed

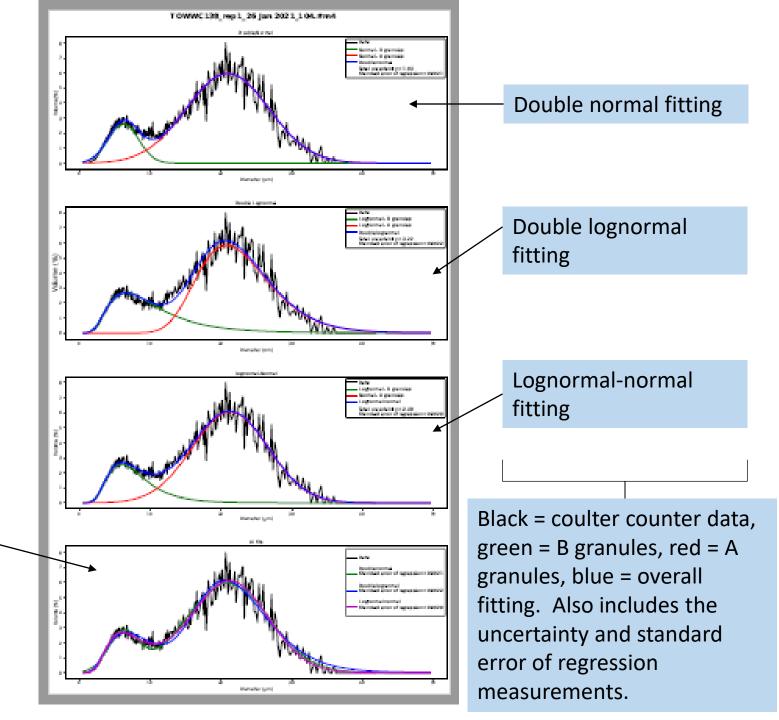
1	A	Successful samples for the double normal fitting	Failed samples for the double normal fitting	D Successful samples for the double lognormal fitting	Failed samples for the double lognormal fitting	F Successful samples for the lognormal- normal fitting	Failed samples for the lognormal-normal fitting
2	0	TOWWC138_rep1_26 Jan 2021_104.#m4	All_successful	TOWWC138_rep1_26 Jan 2021_104.#m4	All_successful	TOWWC138_rep1_26 Jan 2021_104.#m4	All_successful
3	1	TOWWC139_rep1_26 Jan 2021_107.#m4	_	TOWWC139_rep1_26 Jan 2021_107.#m4	_	TOWWC139_rep1_26 Jan 2021_107.#m4	_
4	2	TOWWC140_rep1_26 Jan 2021_111.#m4		TOWWC140_rep1_26 Jan 2021_111.#m4		TOWWC140_rep1_26 Jan 2021_111.#m4	
5	3	TOWWC141_rep1_26 Jan 2021_112.#m4		TOWWC141_rep1_26 Jan 2021_112.#m4		TOWWC141_rep1_26 Jan 2021_112.#m4	
6	4	TOWWC142_rep1_26 Jan 2021_113.#m4		TOWWC142_rep1_26 Jan 2021_113.#m4	*	TOWWC142_rep1_26 Jan 2021_113.#m4	\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \
7	5	TOWWC143_rep1_26 Jan 2021_114.#m4		TOWWC143_rep1_26 Jan 2021_114.#m4		TOWWC143_rep1_26 Jan 2021_114.#m4	
8	6	TOWWC183_rep1_19 Jan 2021_02.#m4	_	TOWWC183_rep1_19 Jan 2021_02.#m4		TOWWC183_rep1_19 Jan 2021_02.#m4	
9							

Tells us if the fitting failed for the double normal (column C), double lognormal (column E) and lognormal-normal (column G), if any of the fittings failed then the filename will be listed in these columns. This means that the model couldn't be fitted to the data most likely as it significantly deviates from being bimodally distributed. It could be the case that one model might not fit but the other two might.

What should I do if the fitting is failing quite often? You could go into the coulter_counter_fitting_script and adjust the initial parameter values to fit your data more closely (HINT – use the visualisation of the initial parameter values to help). If not this could suggest that this is not an appropriate model for your data and you might need to use something else, e.g. is it more unimodal?

PDF files

Comparison of double normal (green), double lognormal (blue) and lognormal-normal (purple) fittings with the raw data (black) and the standard error of regression for each model



These 2 chunks format the script and import all the libraries and packages which are needed.

You'll probably never need to alter these.

```
In [ ]: # Formatting notebook to fit the browser size
        from IPython.core.display import display, HTML
        display(HTML("<style>.container { width:100% !important; }</style>"))
In [ ]: # Loading libraries and packages
        %matplotlib inline
        import numpy as np
        import matplotlib.pyplot as plt
        import matplotlib as mpl
        import matplotlib.ticker as mtick
        import scipy.optimize
        import scipy.stats
        import scipy.integrate as integrate
        import scipy.integrate as special
        import pandas as pd
        import os
        import linecache
        from statistics import mean
        plt.style.use('ggplot') # use gaplot style for graps
```

This is only intended to provide an overview, for more info look at the comments in the script. It is also more for interest/if you might need to alter the script and you can run the script easily without understanding this.

The next chunk imports the data (a CSV file) from the data_inputs folder. It only imports one file at a time.

It sorts and converts all the data to the correct format. Crucially it adjusts the volume for the bin width (column called Diff_volume_density).

```
In [ ]: # Import data
        # path of input file(s) - note that the file(s) has to be in a folder called data inputs
        input data file = os.path.join('data inputs', filename) #note - this uses a variable called filename as the input file, if you are
        #getting the base name of the file which will be used to name the output file in the same way
        name = ((os.path.splitext(input data file))[0]).lstrip('data inputs\\') #This takes the file name, splits it at the extension (e.
        # define column headings for the dataframe
        headings list = ['Bin number', 'Bin diameter lower', 'Diff volume', 'Diff number perc', 'Diff number gram', 'Diff surface area',
        # read in dataframe using pandas library
        input data df = pd.read csv(input data file, sep='\,', engine='python', skiprows=55,
                                    names=headings list)
        # calculate bin width in a new column
        input data df['Bin width'] = input data df['Bin diameter lower'].shift(-1) - input data df['Bin diameter lower']
        # read off the final bin number into a variable, called input data bin max, and then drop this row
        input data bin max = input data df.iloc[-1,1]
        input data df = input data df.drop(input data df.index[input data df.tail(1).index])
        # set 'Bin number' column to be the index
        input data df = input data df.set index('Bin number')
        # change datatype of index from float to integer
        input data df.index = input data df.index.astype('int')
        # Calculate mid-point of bins
        input data df['Bin midpoint'] = input data df['Bin diameter lower']+input data df['Bin width']/2
        # Calc Diff volume density using: Diff volume/Bin width
        input_data_df['Diff_volume_density'] = input_data_df['Diff_volume']/input data_df['Bin_width']
        # convert input data to lists
        bin midpoint input list = input data df['Bin midpoint'].tolist()
        diff volume density input list = input data df['Diff volume density'].tolist()
```

The script defines the lognormal and normal functions mathematically:

```
In []: # Define the Lognormal and normal functions

def lognormal_func(x_, A_1_, mu_1_, sigma_1_):
    return ( (A_1_/(x_*sigma_1_*np.sqrt(2.*np.pi)))*np.exp( -((np.log(x_)-mu_1_)/sigma_1_)**2/2. ) )

def normal_func(x_, A_2_, mu_2_, sigma_2_):
    return ( (A_2_/(sigma_2_*np.sqrt(2.*np.pi)))*np.exp( -( (x_ - mu_2_)/sigma_2_ )**2/2. ) )
```

The script starts with the lognormal-normal fitting:

Defines the lognormal-normal fitting mathematically and makes it into a function, this means that it can be easily used. A_1_, mu_1_ are the parameters which are in the mathematical equation

```
## LOGNORMAL-NORMAL FUNCTION
In [ ]: # Define the lognormal-normal function
       def lognormal_normal_func(x_, A_1_, mu_1_, sigma_1_, A_2_, mu_2_, sigma_2_):
          return (lognormal func(x, A 1, mu 1, sigma 1) + normal func(x, A 2, mu 2, sigma 2))
                                                                               Uses the scipy.optimise.curve fit package to
In []: # Define a function to fit the lognogrmal-normal function _
                                                                               create a function which we will use later to fit
       def func lognormal normal fit(bin midpoint input , diff volume density input ):
                                                                               the lognormal-normal curve to the input data
          #initial parameter values:
          lognormal_initial_params_ = [20.0, 1.8, 0.4] ←
                                                    The initial parameters which are used in the fitting
          normal_initial_params = [50.0, 20.0, 6.0]
          fit params, fit cov = scipy.optimize.curve fit lognormal normal func, bin midpoint input, diff volume density input,
                                                     fit params err = np.sqrt(np.diag(fit cov ))
          return fit params , fit params err
                                                                               We restrict the values of the parameters (e.g. so
       #This function will return: {list of fit parameters}, {list of fit uncertanties} we don't get negative diameters) using bounds
```

How does scipy.optimisie.curve_fit actually perform the fitting? It uses a non-linear least squares method

```
# #Plots the initial parameters to visualise how close the initial parameters fit
# #This can be useful if your curves are very different from normal and the initial parameters need to be adjusted
# #Commented out so not produced when running in batch mode
                                                                            Make sure these match the
# #Initial parameter estimates - ensure these are the same as shown above
# lognormal_initial_params = [20.0, 1.8, 0.4] ___
                                                                            initial parameters in the
# normal initial params = [50.0, 20.0, 6.0]
                                                                            function in the chunk
# # define x values to plot
\# \times \text{vals plot} = \text{np.aranae}(0.5,50.0.5)
                                                                            above
# # define curves to fit
# lognormal_vals_curve = [lognormal_func(i_, lognormal_initial_params[0],log.....___
                                        lognormal initial params[2]) for i in x vals plot]
# normal vals curve = [normal func(i , normal initial params[0],normal initial params[1],normal initial params[2])
                      for i in x vals plot]
# lognormal_normal_vals_curve = [lognormal_normal_func(i_, lognormal_initial_params[0],lognormal_initial params[1],lognormal_init
                                        normal initial params[0], normal initial params[1], normal initial params[2])
                         for i in x vals plot]
# #plot graphs of initial parameters
# fig, ax = plt.subplots(1,1,figsize=(15,8))
# for spine in ['left', 'right', 'top', 'bottom']:
     ax.spines[spine].set color('k')
# ax.set facecolor('white')
# ax.plot(input data df['Bin midpoint'].tolist(), input data df['Diff volume density'].tolist(),
         linewidth=2, color='k',label='Data' )
# ax.plot(x_vals_plot, lognormal_vals_curve,
         linewidth=2, color='green',label='Lognormal' )
# ax.plot(x_vals_plot, normal_vals_curve,
         linewidth=2, color='r',label='Normal' )
# ax.plot(x vals plot, lognormal normal vals curve,
         linewidth=2, color='b',label='Lognormal-normal' )
# fig.suptitle('Initial parameter guess: '+name, fontsize=16)
# ax.set xlabel("Diameter ($\mu $m)", fontsize=18)
# ax.xaxis.set major locator(plt.MaxNLocator(6))
# ax.set ylabel("Volume (%)", fontsize=18)
# ax.tick_params(axis='both', which='major', labelsize=16)
# leg 00 = ax.legend(fontsize=16, loc='upper right', ncol=1,facecolor='white', framealpha=1)
# leg 00.get frame().set edgecolor('k')
# plt.tight_layout()
# fig.subplots_adjust(top=0.9)
```

Optional - Sometimes we might want to visualise how the initial curve (using the initial parameters) fits the data. This chunk will plot the initial curve over the input_data

This script has already been optimised for bimodal distributions, but this chunk could be useful if you want to change the initial parameters to fit traces which look substantially different (if you've already checked to make sure the current model isn't fitting properly).

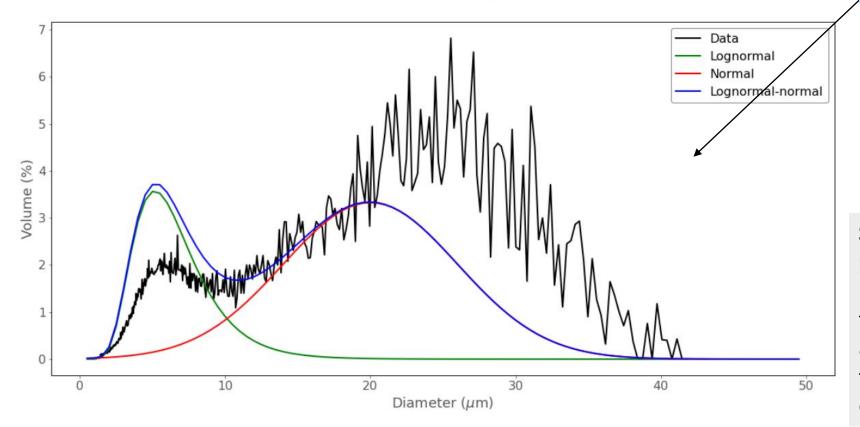
At the moment this chunk won't run as it has been commented out, to uncomment it select all the text and press CTRL and /

The output graph will be produced at the bottom of the Jupyter notebook (in the Running_script) and NOT as a PDF.

How does the coulter_counter_fitting_script work? The initial curve doesn't

Example output from the initial parameter plot:

Initial parameter guess: TOWWC003_rep1_26 Jan 2021_03.#m4



The initial curve doesn't have to be that close in shape to the final curve, this example still produces a successful fitting.

Should I bother to plot the initial parameters? Personally I wouldn't bother doing this as I've found that the initial parameters are fine and allow the model to be fitted to a wide range of bimodal coulter counter traces.

The next chunk tries to fit the lognormal-normal fitting to the input data:

```
There is an exception catcher – this means
In [ ]: # Run the lognormal-normal fitting
                                                                                                that if the model can't be fitted to the input
       #create empty lists which the successful and non-successful fittings will be stored in
       failed lognormal normal = []
                                                                                                data the error won't cause the script to
       successful lognormal normal = []
                                                                                                crash. Instead all the parameters will be
       fitting failed lognormal normal = []
                                                                                                saved as 'fitting failed'
       #run the lognormal-normal fit with the exception catcher
        try:
           output_lognormal_normal_fit_params_list, output_lognormal_normal_fit_err = func_lognormal_normal_fit(bip_midpoint input list.
       exc We are using the function we defined earlier with the
                                                                                 get \n included in the sample names when we make a
            scipy.optimise.curve fit package
               #this sets the parameters to fitting failed so in the output file we get this message
               B granule diameter lognormal normal = "fitting failed"
               A granule diameter lognormal normal = "fitting failed"
               B granule area lognormal normal = "fitting failed"
               A granule area lognormal normal = "fitting failed"
               B granule content lognormal normal = "fitting failed"
               lognormal_normal_uncertainity = "fitting failed"
               lognormal normal standard error of regression = "fitting failed" -
               #this is used in an if statement during plotting so that if failed we don't get a graph
               fitting failed lognormal normal = "yes"
```

Also in this chunk we calculate the granule diameters, granule content, uncertainty and standard error of regression:

```
successful lognormal normal = name.split('\n') #we need the \n if not we get \n included in the sample names when we make a
output lognormal fit params list = output lognormal normal fit params list[:3]
output normal fit params list = output lognormal normal fit params list[3:]
#Getting the parameters from the models:
#lognorm mean is equal to exp(mu + (sigma^2/2))
B granule diameter lognormal normal = np.exp(output lognormal normal fit params list[1] + ((output lognormal normal fit param
#norm mean is equal to mu
A granule diameter lognormal normal = output lognormal normal fit params list[4]
#areas under curves - with the scipy integrate function: the individual functions have been redefined using the output parame
#and the integrate function is used, the output is a tuple with an estimated value of the integral first and the second value
#B granules - lognorm fitting
def lognorm func B granules(x ):
return ( (output_lognormal_normal_fit_params_list[0]/(x_output_lognormal_normal_fit_params_list[2]*np.sqrt(2.*np.pi)))*
B_granule_area_lognormal_normal = scipy.integrate.quad ognorm_func_B_granules,0, np.inf) #integrate between 0 and infinity
#A granules - normal fitting
def normal func A granules(x ):
        return ( (output_lognormal_normal_fit_params_list[3]/(output_lognormal_normal_fit_params_list[5]*np.sqrt(2.*np.pi)))*np
A granule area lognormal normal = scipy.integrate.quad(normal func A granules, 0, np.inf) #integrate between 0 and inf
Total area lognormal normal = A granule area lognormal normal[0] + B granule area lognormal normal[0]
B granule content lognormal normal = B granule area lognormal normal[0]/Total area lognormal
#reduced uncertainity
lognormal normal uncertainity = np.sum(output lognormal normal fit err)
#standard error of regression
lognormal normal vals list = [lognormal normal func(input data df['Bin midpoint'].tolist()[i], *output lognormal normal fit
                                        for i in range(len(input data df['Diff volume density']))]
lognormal_normal_standard_error_of_regression = np.sqrt( (1/len(input_data_df['Diff_volume_density']-2)) * (sum((np.array(ing.array)) * (sum((np.array)) * (sum((np.array)) * (sum((np.array))) * (sum((np.array)) * (sum((np.array))) * (sum((np.arra
```

The diameters are calculated from the means of the mathematical equations

The scipy.integrate.quad package is used to integrate the area under the curves and the relative areas are used to calculate the granule content

The uncertainty is the sum of the errors of all the individual parameters

Standard error of regression, uses this equation: $s(b_1) = \sqrt{\frac{1}{n-2} * \sum_{i=0}^{n} (y_i - \hat{y}_i)}$

- **n**: total sample size
- y_i: actual value of response variable
- $\hat{\boldsymbol{y}}_{i}\!\!:$ predicted value of response variable
- \boldsymbol{x}_i : actual value of predictor variable
- x̄: mean value of predictor variable

Plotting the final fitting:

work?

This is commented out as we want to save it in a PDF and not just display the final fitting graph here.

```
In [ ]: # # Plot optimized fit of lognormal-normal fitting (if it was successful)
        # # Commented out so not produced when running in batch mode
        # if fitting failed lognormal normal == "yes":
        # else:
              # define x values to plot
              x vals plot = np.arange(0.5,50,0.5)
              # define curves to fit
              lognormal vals curve = [lognormal func(i , *output lognormal fit params list) for i in x vals plot]
              normal_vals_curve = [normal_func(i , *output_normal_fit_params_list) for i _in x_vals_plot]
              lognormal normal vals curve = [lognormal normal func(i , *output lognormal normal fit params list) for i in x vals plot]
              #plot graphs of initial parameter
              fig, ax = plt.subplots(1,1,figsize=(15,8))
              for spine in ['left', 'right', 'top', 'bottom']:
                  ax.spines[spine].set color('k')
              ax.set facecolor('white')
              ax.plot(input data df['Bin midpoint'].tolist(), input data df['Diff volume density'].tolist(),
                      linewidth=2, color='k',label='Data' )
              ax.plot(x vals plot, lognormal vals curve,
                      linewidth=2, color='green', label='Lognormal - B granules' )
              ax.plot(x vals plot, normal vals curve,
                     linewidth=2, color='r', label='Normal - A granules' )
              ax.plot(x vals plot, lognormal normal vals curve,
                      linewidth=2, color='b', label='Lognormal-normal')
              ax.plot(x vals plot, lognormal vals curve,
                      linewidth=2, color='w', alpha=0, label='Total uncertainty = %.2f\nStandard error of regression =%.4f' %
                     (lognormal normal uncertainity, lognormal normal standard error of regression))
              fig.suptitle('Lognormal-normal fit: '+name, fontsize=16)
              ax.set xlabel("Diameter ($\mu $m)", fontsize=18)
              ax.xaxis.set major locator(plt.MaxNLocator(6))
              ax.set ylabel("Volume (%)", fontsize=18)
              ax.tick params(axis='both', which='major', labelsize=16)
              leg 00 = ax.legend(fontsize=16, loc='upper right', ncol=1, facecolor='white', framealpha=1)
              leg 00.get frame().set edgecolor('k')
              plt.tight layout()
              fig.subplots adjust(top=0.9)
```

- The script goes through the same process with the double normal and then the double lognormal fittings
- It then combines the fits together in one graph:

```
In [ ]: # # Plotting all fits
        # # Commented out so not produced when running in batch mode
        # #defining x values
        \# \times \text{vals plot} = \text{np.arange}(0.5,50,0.5)
        # fig, ax = plt.subplots(1,1,figsize=(15,8))
        # for spine in ['left', 'right', 'top', 'bottom']:
        # ax.spines[spine].set_color('k')
        # ax.set_facecolor('white')
        # ax.plot(input_data_df['Bin_midpoint'].tolist(), input_data_df['Diff_volume_density'].tolist(),
                  linewidth=2, color='k',label='\nData\n' )
        # if fitting failed double normal == "yes":
        # else:
            ax.plot(x_vals_plot, double_normal_vals_curve,
                  linewidth=2, color='g',label='Double normal\nStandard error of regression =%.4f\n' % (double normal standard error of r
        # if fitting failed double lognormal == "yes":
              ax.plot(x_vals_plot, double_lognormal_vals_curve,
                  linewidth=2, color='b',label='Double lognormal\nStandard error of regression =%.4f\n' % (double lognormal standard error
        # if fitting failed lognormal normal == "yes":
        # eLse:
             ax.plot(x_vals_plot, lognormal_normal_vals_curve,
                  linewidth=2, color='m',label='Lognormal-normal\nStandard error of regression =%.4f\n' % (lognormal normal standard error
        # ax.set xlabel("Diameter ($\mu $m)", fontsize=18)
        # ax.xaxis.set major locator(plt.MaxNLocator(6))
        # ax.set_ylabel("Volume (%)", fontsize=18)
        # ax.tick_params(axis='both', which='major', labelsize=16)
        # ax.set_title('All fits', fontsize=18)
        # leg 00 = ax.legend(fontsize=16, loc='upper right', ncol=1, facecolor='white', framealpha=1)
        # leg 00.get frame().set edgecolor('k')
```

This is commented out as we want to save it in a PDF and not just display the final fitting graph here.

This chunk then plots four graphs 1) Double normal fitting, 2) Double lognormal fitting, 3) Lognormal-Normal fitting, 4) Overlay of all 3 fittings and saves it as a PDF

```
# Combining all plots together in one pdf file
# Note - the if-else loops are used so that if the fitting failed then the graph is blank and says fitting failed
# Note - saves the pdf in a folder called output

#defining x values
x_vals_plot = np.arange(0.5,50,0.5)

fig, ax = plt.subplots(4,1,figsize=(15,25))
fig.suptitle(name, fontsize=20, y=1.01, fontweight="bold")
fig.subplots_adjust(top=0.95)
fig.tight_layout(h_pad=8)

#first graph - double normal graph
```

Long chunk so not all shown here

```
#Saving plots in a folder called output
plot_name = (name.strip('\n')+"_output.pdf")
fig.savefig('output/'+name+".pdf", bbox_inches = 'tight', dpi=300)
```

The final chunk saves all the parameters from the different fittings and saves them to a variable called 'fitting_parameters' which the Batch_running_of_coulter_counter_script uses

```
#Getting the parameters from the different fits

Double_normal_parameters = (B_granule_diameter_double_normal, A_granule_diameter_double_normal, B_granule_content_double_normal,

Double_lognormal_parameters = (B_granule_diameter_double_lognormal, A_granule_diameter_double_lognormal, B_granule_content_double_lognormal_normal_parameters = (B_granule_diameter_lognormal_normal, A_granule_diameter_lognormal_normal, B_granule_content_lognor

#Joining together the sample name and fitting parameters

name_str = (name,) #converts the sample name into a tuple so it is in the same format as the parameters

fitting_parameters = (name_str + Double_normal_parameters + Double_lognormal_parameters + Lognormal_normal_parameters)
```

Starts the same as the coulter_counter_fitting script

```
#Running in batch mode
#This needs a folder called data inputs with csv files to use and a folder called output which it will put the fitting parameters
parameters from fittings = [] #need to make an empty data frame with which I can add the fitting parameters to
#empty lists which I will fill with sample names which have either worked or failed
did not run = []
failed lognormal normal fit list = []
successful lognormal normal fit list = []
failed double normal fit list=[]
successful double normal fit list=[]
failed_double_lognormal_fit_list=[]
successful double lognormal fit list=[]
#this is a for loop which reads input files in a directory called data inputs and years them as input for the coulter counter fit
for filename in os.listdir('data inputs/'): #runs with all files in the data imputs folder
   if filename.endswith(".csv") or filename.endswith(".CSV"): #will only work on csv files, so means you can have other file type
        file name = filename #file name is the variable which the run coulter count fit rose using python fitting script uses as
            %run coulter counter fitting script.ipynb #this line runs the script
        except IndexError or RuntimeError:
            did not run.append(filename)
        else:
            parameters_from_fittings.append(fitting_parameters) #takes the fitting parameters from all the samples and stores tog
            #creating lists of samples which have been successfut and those which have failed
            successful lognormal normal fit list.append(successful lognormal normal)
            failed lognormal normal fit list.append(failed lognormal normal)
            successful double normal fit list.append(successful double normal)
            failed double normal fit list.append(failed double normal)
            successful double lognormal fit list.append(successful double lognormal)
            failed double lognormal fit list.append(failed double lognormal)
    continue
```

It reads through all CSV files in the data_inputs folder and one by one runs them with the coulter_counter_fitting_script

If there is an error it stores the filename but it shouldn't cause the script to crash

The fitting_parameters variable (which contains all the parameters from the different fittings) are saved in one large data frame

```
#concatenates the fitting parameters into an excel file
#produces an output called 'Parameters_from_fittings.xlsx'
parameters_from_fittings_data_frame = pd.DataFrame(parameters_from_fittings, columns=['Sample_name','B granule diameter double no
                                                                                       'B granule content double normal fitting',
                                                                                       'Standard error of regression for the doub!
                                                                                       'A granule diameter double lognormal fittir
                                                                                       'Uncertainity for the double logno
                                                                                       'B granule diameter logno
                                                                                       'B granule content lognormal-normal fitting
                                                                                        Standard error of regression for the logno
parameters from fittings data frame.to excel("output/Parameters from fittings.xlsx")
#creating data frames with the fittings which were successful and those which failed
successful lognormal normal fit dataframe = pd.DataFrame(successful lognormal normal fit list)
failed lognormal normal fit dataframe = pd.DataFrame(failed lognormal normal fit list)
successful double lognormal fit dataframe = pd.DataFrame(successful double lognormal fit list)
failed double lognormal fit dataframe = pd.DataFrame(failed double lognormal fit list)
successful double normal fit dataframe = pd.DataFrame(successful double normal fit list)
failed_double_normal_fit_dataframe = pd.DataFrame(failed_double_normal_fit_list)
#if loops which (if the dataframes are empty) fills it with either all successful or all failed as appropriate
All successful="All successful"
All failed = "All failed"
if failed lognormal normal fit dataframe.empty:
   failed lognormal normal fit dataframe = pd.DataFrame(['All successful'])
if failed double lognormal fit dataframe.empty:
   failed_double_lognormal_fit_dataframe = pd.DataFrame(['All_successful'])
if failed double normal fit dataframe.empty:
   failed double normal fit dataframe = pd.DataFrame(['All successful'])
if successful lognormal normal fit dataframe.empty:
   successful lognormal normal fit dataframe = pd.DataFrame(['All failed'])
if successful double lognormal fit dataframe.empty:
   successful double lognormal fit dataframe = pd.DataFrame(['All failed'])
if successful double normal fit dataframe.empty:
   successful double normal fit dataframe = pd.DataFrame(['All failed'])
#merging the data frames together and saving the results in excel
"#produces an output called 'Samples which fitted successfully and samples which failed.xlsx'
successful and failed fittings = pd.concat([successful double normal fit dataframe[0], failed double normal fit dataframe[0], suc
                    failed double lognormal fit dataframe[0], successful lognormal normal fit dataframe[0], failed lognormal norm
successful and failed fittings results = successful and failed fittings.set axis(["Successful samples for the double normal fit
                                         "Successful samples for the double lognormal fitting", "Failed samples for the doubl
                                         "Successful samples for the lognormal-normal fitting", "Failed samples for the lognormal
successful and failed fittings results to excel("output/Samples which fitted successfully and samples which failed.xlsx")
```

The fitting parameters are all saved together in one Excel file

Creates an Excel file which contains which fittings worked/didn't work for each file – called

"Samples_which_fitted_successfully_and _samples_which_failed"

Creates an Excel file which contains the list of filenames which didn't import properly, this is called "Samples_where_the_script_failed"

```
#creating a file of samples which failed - most likely during data import
did_not_run_dataframe = pd.DataFrame(did_not_run, columns=["Samples where the script failed - most likely during data import, so
did_not_run_dataframe.to_excel("output/Samples_where_the_script_failed.xlsx")
```

Overview

1) You submit it using the Running script

You only need to do this first step, everything else is automated

- 2) The Running script starts the Batch_running_of_coulter_counter_script
- 3) The Batch_running_of_coulter_counter_script takes every csv file in the data_inputs folder and submits it to the coulter_counter_fitting_script (one at a time)
- 4) The coulter_counter_fitting_script performs the fitting