

# MainMethod

March 25, 2021

Taking the output from DEG-SEQ2, the data “5LS\_L2L3Combined.csv” contains the 5 life stages we are interested in: Embryo, L1 larva, Dauer Larva, L2L3 Larva and Adult, lets take a peek of that data

```
[1]: import csv
import os

#user configurable variables
number_of_lines_to_print=10
expressionCountFile=os.path.join(os.getcwd(),'csvs/5LS_L2L3Combined.csv')
#Code Chunk for printing the file
with open(os.path.join(os.getcwd(),expressionCountFile)) as csv_file:
    csv_reader = csv.reader(csv_file, delimiter=',')
    for row in csv_reader:
        print(row)
        number_of_lines_to_print-=1
        if number_of_lines_to_print<=0:
            break
```

```
['WBID', 'elongating embryo Ce', 'L1 larva Ce', 'dauer larva Ce', 'adult Ce',
'L2L3_larva']
['WBGene000000001', '4208', '12140', '5547', '2246', '2369']
['WBGene000000002', '12554', '7828', '831', '280', '2591']
['WBGene000000003', '7180', '11253', '570', '212', '2466']
['WBGene000000004', '33305', '26947', '3212', '576', '5391']
['WBGene000000005', '595', '132', '37', '281', '1410']
['WBGene000000006', '425', '12243', '3146', '228', '2446']
['WBGene000000007', '36', '314', '129', '197', '1719']
['WBGene000000008', '0', '19', '663', '19', '182']
['WBGene000000009', '71', '416', '193', '20', '64']
```

Lets look at some statistics about the data:

```
[2]: import pandas as pd
import numpy as np

exp_data = pd.read_csv(expressionCountFile)
```

```
print(exp_data.describe())
```

	elongating embryo Ce	L1 larva Ce	dauer larva Ce	adult Ce \
count	20361.000000	2.036100e+04	2.036100e+04	2.036100e+04
mean	3692.118364	1.097686e+04	3.908223e+03	2.065875e+03
std	12796.637118	5.382926e+04	2.111976e+04	2.269343e+04
min	0.000000	0.000000e+00	0.000000e+00	0.000000e+00
25%	8.000000	5.700000e+01	2.700000e+01	5.000000e+00
50%	201.000000	7.240000e+02	4.030000e+02	8.600000e+01
75%	2730.000000	4.645000e+03	2.375000e+03	8.690000e+02
max	355180.000000	1.890193e+06	1.303599e+06	2.253663e+06

	L2L3_larva
count	2.036100e+04
mean	4.178275e+03
std	1.964655e+04
min	0.000000e+00
25%	2.900000e+01
50%	3.910000e+02
75%	2.292000e+03
max	1.103229e+06

Now, we need to determine the genes that we consider to be life stage biased, here are the some criterias that must be fullfilled to be considered a life stage biased gene:

This gene has the highest expression in that life stage

This gene's expression at this life stage has at least a fold difference of 2 comparing the max expression in other life stages

At least one life stage has a count that is higher than at least 10% of of counts across all life stages.  
 \*This ensures we dont include genes that have high fold diff due to unbalanced low expression counts, for example, a gene has a count of 1 in one life stage and are not found in other life stages(0 counts),this gene is a uniformly lowly expressed gene in all life stages, however, using the criteria one, this gene would have a fold difference of infinity, by setting a lower bound filter, we exclude these extremely lowly expressed gene counts that are prone to sequencing uncertainties.

Let's process the expression file using above criterias:

```
[3]: from Code import LifeStageBiased as LSB
     #Speficy input and output
     LSB.inputFile= expressionCountFile
     outputPath=os.path.join(os.getcwd(), 'csvs/LSB.csv')
     LSB.outputFile= outputPath
     LSB.cutLowPercentile=0.15
     LSB.foldDiff=2
     LSB.fixedCutValue=0 #This overrides the percentil cut value, set to 0 disables
     ↪ it
     LSB.main()
```

The cutOff Value for the specified percentaile is: 5.0

\*In the data we are analyzing, there are very few genes that have observed expression only in one life stage, in which case the max expression for other life stages is 0, this will yield infinity for the fold diff value, in theory, these are “life stage specific genes” rather than “life stage biased genes”, however, since the same reason we mentioned above about the sequencing uncertainties, we do not believe that a gene with a few counts only in one life stage is more likely to be a life stage biased gene comparing to a gene with high expression in one life stage and very low expression in other life stages, see example below:

```
[4]: ls_data = pd.read_csv(outputFilePath)

print(ls_data.loc[ls_data['GeneID'] == "WBGene00015845"])
print()
print(ls_data.loc[ls_data['GeneID'] == "WBGene00000609"])
```

	GeneID	LS	LS_EXP	SecondMax	RestMean	FoldDiff
6043	WBGene00015845	adult Ce	7.0	0.0	0.0	7.0

	GeneID	LS	LS_EXP	SecondMax	RestMean	FoldDiff
218	WBGene00000609	adult Ce	71952.0	148.0	76.75	486.162162

As shown above, Gene “WBGene00015845” is a relatively lowly expressed gene that is only expressed in adult stage, and “WBGene00000609” has significantly higher expression in adult stage comparing to other life stages, which one are we more confident to select as the life stage biased gene?

There is no sure way to know, to compensate that, instead of putting infinity as the fold difference value for these “life stage specific” genes and put more confidence in them above all other genes, we decided to use their expression value as their foldDiff value, in which case a highly expressed “life stage specific” gene will be given higher confidence comparing to a lowly expressed one.

We are aware that this is perhaps not the best way of handling these genes, but luckily, there are only 15 such genes out of the 10099 life stage biased genes(0.15%) we selected using above filter, so it is extremely unlikely that different handlings of these genes will make a significant difference.

Now the genes that fit into our criteria should be in the *outputFilePath* we set ealier, lets take a look at some basic statistics of these selected life stage biased genes:

```
[5]: ls_data.head()
print(ls_data.columns)

ls_data[["LS_EXP", "SecondMax", "FoldDiff"]]=ls_data[["LS_EXP", "SecondMax", "FoldDiff"]].
    ↪apply(pd.to_numeric)
ls_data.sort_values(by=['FoldDiff'],ascending=False)
ls_data['FoldDiff_RestMean']=ls_data["LS_EXP"]/ls_data["RestMean"]
ls_data['LS_EXP_LOG']=np.log((ls_data['LS_EXP']))
ls_data['FoldDiff_LOG']=np.log((ls_data['FoldDiff']))
ls_data['RestMean_LOG']=np.log((ls_data['RestMean']))
ls_data["FoldDiff_RestMean_LOG"]= np.log(ls_data["FoldDiff_RestMean"])
```

```

print(ls_data.describe())
print()
print("ls_data Summary By Life Stage Group")
print(ls_data.groupby("LS").describe())

ax=ls_data['LS_EXP_LOG'].hist(by=ls_data['LS'],range=[0,12])

```

```

Index(['GeneID', 'LS', 'LS_EXP', 'SecondMax', 'RestMean', 'FoldDiff'],
      dtype='object')

```

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	1.009900e+04	10099.000000	10099.000000	10099.000000
mean	1.679783e+04	4102.327656	2191.947742	15.490207
std	7.351532e+04	19035.148534	10168.396260	176.069930
min	5.000000e+00	0.000000	0.000000	2.000000
25%	1.560000e+02	28.000000	12.250000	2.617371
50%	1.298000e+03	248.000000	108.750000	3.776471
75%	8.224500e+03	1946.000000	988.500000	6.722003
max	2.253663e+06	540124.000000	315851.750000	12412.000000

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG \
count	1.009900e+04	10099.000000	10099.000000	1.009900e+04
mean	inf	7.073785	1.586584	-inf
std	NaN	2.538669	0.907085	NaN
min	2.181818e+00	1.609438	0.693147	-inf
25%	5.582012e+00	5.049856	0.962170	2.505526e+00
50%	8.444388e+00	7.168580	1.328790	4.689052e+00
75%	1.537575e+01	9.014873	1.905386	6.896189e+00
max	inf	14.628067	9.426419	1.266303e+01

	FoldDiff_RestMean_LOG
count	1.009900e+04
mean	inf
std	NaN
min	7.801586e-01
25%	1.719549e+00
50%	2.133502e+00
75%	2.732792e+00
max	inf

ls\_data Summary By Life Stage Group

	LS_EXP					
	count	mean	std	min	25%	
LS						
L1 larva Ce	5426.0	25057.749355	89726.448364	5.0	187.00	
L2L3_larva	1018.0	4984.386051	13889.489577	5.0	124.00	
adult Ce	695.0	13820.099281	111338.218970	5.0	131.00	

dauer larva Ce	1550.0	3776.957419	15036.754400	5.0	83.00
elongating embryo Ce	1410.0	9322.402128	25125.246231	5.0	517.75

	50%	75%	max	SecondMax count	mean	...	\
LS						...	
L1 larva Ce	2159.5	14543.50	1890193.0	5426.0	6524.922226	...	
L2L3_larva	547.0	2903.50	173036.0	1018.0	1200.904715	...	
adult Ce	619.0	3951.00	2253663.0	695.0	842.952518	...	
dauer larva Ce	300.0	1844.00	235883.0	1550.0	691.027097	...	
elongating embryo Ce	2472.5	7250.25	328012.0	1410.0	2231.002837	...	

	RestMean_LOG 75%	FoldDiff_RestMean_LOG max	count	mean	std	\
LS						
L1 larva Ce	7.677081	12.663028	5426.0	inf	NaN	
L2L3_larva	5.468584	10.355311	1018.0	inf	NaN	
adult Ce	5.523903	9.268963	695.0	inf	NaN	
dauer larva Ce	4.901099	10.365506	1550.0	inf	NaN	
elongating embryo Ce	6.927496	10.556119	1410.0	inf	NaN	

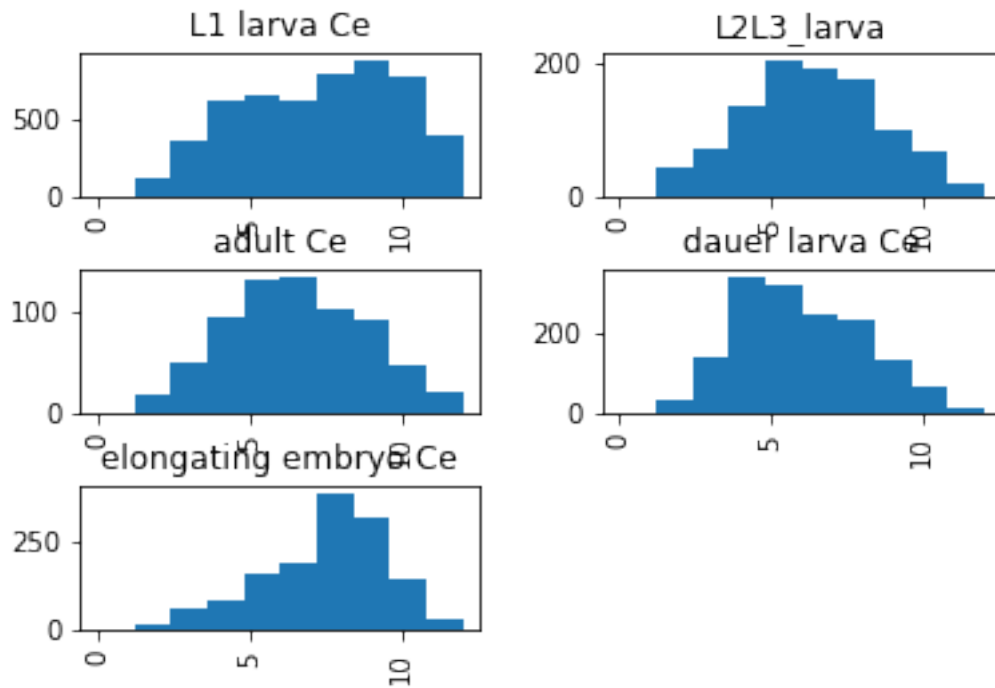
	min	25%	50%	75%	max
LS					
L1 larva Ce	0.837757	1.684538	2.082542	2.593047	inf
L2L3_larva	0.894745	1.806794	2.172522	2.710890	inf
adult Ce	0.856746	1.833032	2.234532	2.773614	inf
dauer larva Ce	0.822815	1.857852	2.420250	3.399912	inf
elongating embryo Ce	0.780159	1.654225	2.079442	2.765388	inf

[5 rows x 72 columns]

/home/lu/.local/lib/python3.8/site-packages/pandas/core/series.py:726:

RuntimeWarning: divide by zero encountered in log

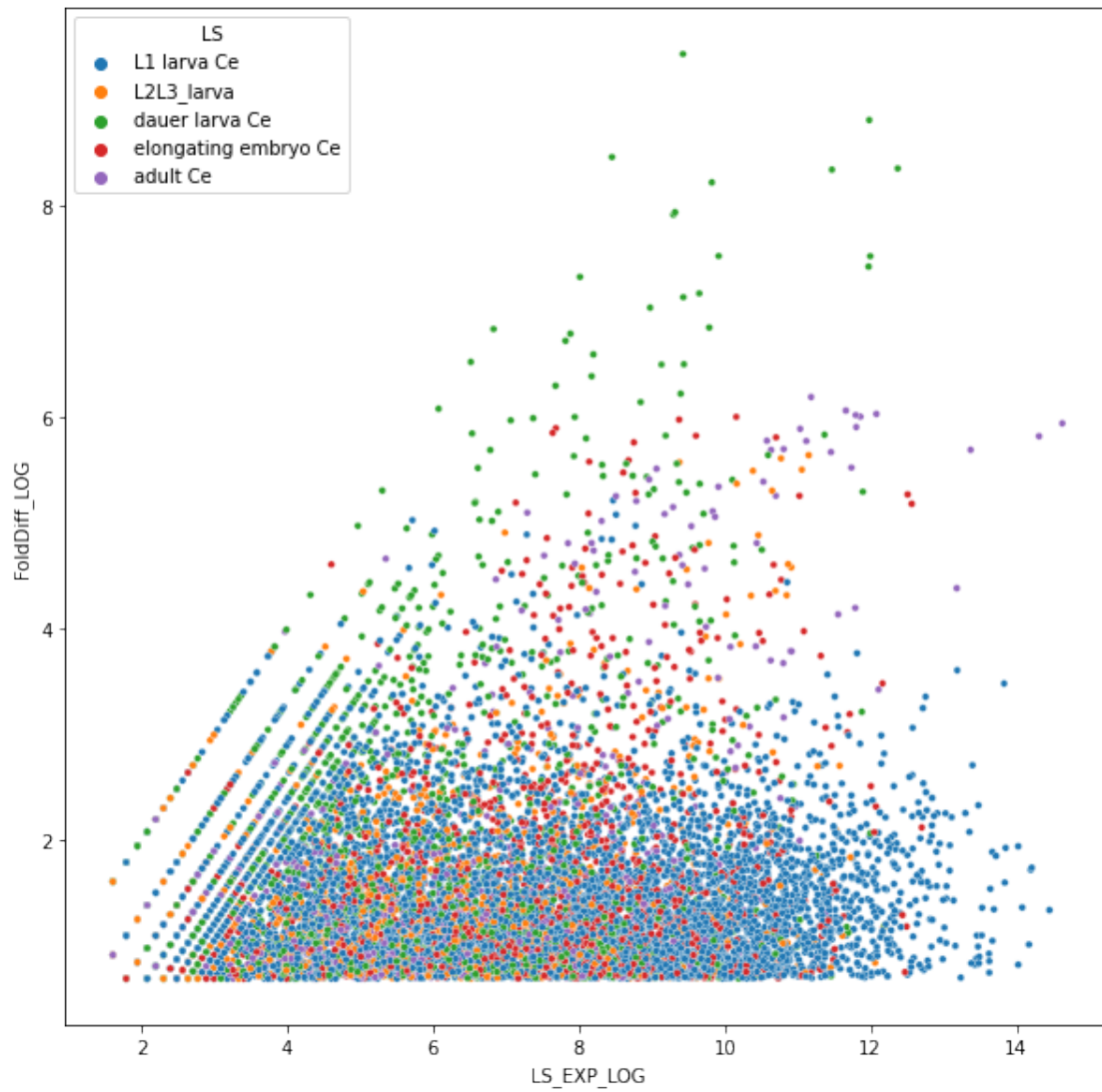
result = getattr(ufunc, method)(\*inputs, \*\*kwargs)



```
[6]: from matplotlib import pyplot as plt
import seaborn as sns

ax2=plt.figure(figsize=[10,10])
sns.scatterplot(x='LS_EXP_LOG',y='FoldDiff_LOG',hue='LS', data=ls_data,s=15)
```

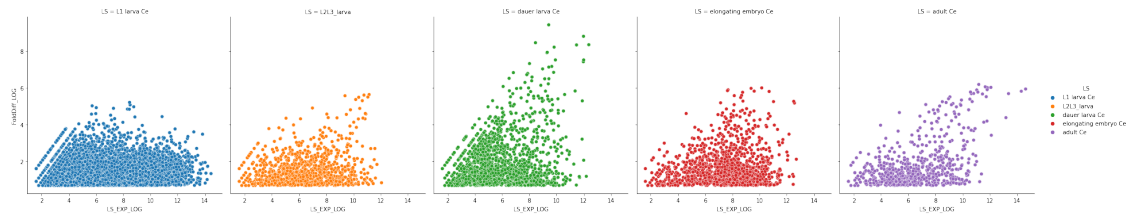
```
[6]: <AxesSubplot:xlabel='LS_EXP_LOG', ylabel='FoldDiff_LOG'>
```



```
[7]: ax3=plt.figure(figsize=[20,20])
sns.relplot(
    data=ls_data,x='LS_EXP_LOG', y="FoldDiff_LOG",
    col="LS", hue="LS",
    kind="scatter"
)
```

[7]: <seaborn.axisgrid.FacetGrid at 0x7f67fac79070>

<Figure size 1440x1440 with 0 Axes>



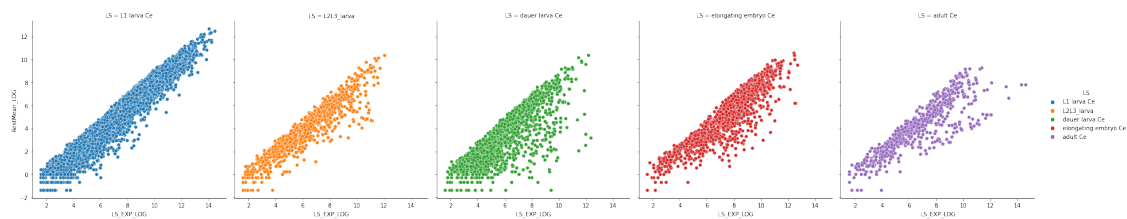
[ ]:

Look at the relationship between the max expression vs mean of expression in other life stages

```
[8]: ax4=plt.figure(figsize=[20,20])

sns.relplot(
    data=ls_data,x='LS_EXP_LOG', y="RestMean_LOG",
    col="LS", hue="LS",
    kind="scatter"
)
```

<Figure size 1440x1440 with 0 Axes>



Now, lets look at the number of genes from each life stage selected when we change the threshold:

```
[9]: sorted_ls_data=ls_data.sort_values(['LS','FoldDiff'],ascending=False)

thresholds=[2**i for i in range(1,11)]

for threshold in thresholds:
    df_filtered=sorted_ls_data.loc[sorted_ls_data['FoldDiff'] >= threshold]
    df_count=df_filtered.groupby("LS").count()
    ax=plt.figure(figsize=[8,6])
    text=("Threshold of FoldDiff: "+ str(threshold))
    sns.histplot(df_filtered, x="LS",hue="LS").set_title(text)
    print(text)
    print(df_filtered.describe())
```



Threshold of FoldDiff: 2

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	1.009900e+04	10099.000000	10099.000000	10099.000000
mean	1.679783e+04	4102.327656	2191.947742	15.490207
std	7.351532e+04	19035.148534	10168.396260	176.069930
min	5.000000e+00	0.000000	0.000000	2.000000
25%	1.560000e+02	28.000000	12.250000	2.617371
50%	1.298000e+03	248.000000	108.750000	3.776471
75%	8.224500e+03	1946.000000	988.500000	6.722003
max	2.253663e+06	540124.000000	315851.750000	12412.000000

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG \
count	1.009900e+04	10099.000000	10099.000000	1.009900e+04
mean	inf	7.073785	1.586584	-inf
std	NaN	2.538669	0.907085	NaN
min	2.181818e+00	1.609438	0.693147	-inf
25%	5.582012e+00	5.049856	0.962170	2.505526e+00
50%	8.444388e+00	7.168580	1.328790	4.689052e+00
75%	1.537575e+01	9.014873	1.905386	6.896189e+00
max	inf	14.628067	9.426419	1.266303e+01

	FoldDiff_RestMean_LOG
count	1.009900e+04
mean	inf
std	NaN
min	7.801586e-01
25%	1.719549e+00
50%	2.133502e+00
75%	2.732792e+00
max	inf

Threshold of FoldDiff: 4

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	4.736000e+03	4736.000000	4736.000000	4736.000000
mean	1.874109e+04	2476.493454	1307.303262	29.896106
std	8.095115e+04	10934.194589	5513.457001	256.362229
min	5.000000e+00	0.000000	0.000000	4.000000
25%	1.450000e+02	15.000000	6.500000	5.108633
50%	1.014000e+03	95.000000	41.625000	7.072728
75%	7.755500e+03	829.250000	387.875000	12.500000
max	2.253663e+06	261525.000000	118162.250000	12412.000000

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG \
count	4736.000000	4736.000000	4736.000000	4736.000000
mean	inf	7.027185	2.253014	-inf
std	NaN	2.558391	0.934246	NaN
min	4.428379	1.609438	1.386294	-inf
25%	10.942227	4.976734	1.630932	1.871802
50%	16.101432	6.921658	1.956246	3.728696

75%	29.341025	8.956157	2.525729	5.960682
max	inf	14.628067	9.426419	11.679814

	FoldDiff_RestMean_LOG
count	4736.000000
mean	inf
std	NaN
min	1.488034
25%	2.392629
50%	2.778908
75%	3.378987
max	inf

Threshold of FoldDiff: 8

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	2.065000e+03	2065.000000	2065.000000	2065.000000
mean	1.737960e+04	998.086199	542.603995	61.463707
std	8.397990e+04	4394.859318	2509.116448	386.007561
min	8.000000e+00	0.000000	0.000000	8.000000
25%	1.510000e+02	8.000000	3.750000	10.000000
50%	9.560000e+02	40.000000	19.500000	14.112252
75%	6.565000e+03	291.000000	129.250000	27.111111
max	2.253663e+06	69368.000000	34733.500000	12412.000000

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG \
count	2065.000000	2065.000000	2065.000000	2065.000000
mean	inf	6.996054	2.989515	-inf
std	NaN	2.438047	0.995191	NaN
min	8.000000	2.079442	2.079442	-inf
25%	22.040000	5.017280	2.302585	1.321756
50%	32.920635	6.862758	2.647043	2.970414
75%	64.631579	8.789508	3.299944	4.861749
max	inf	14.628067	9.426419	10.455460

	FoldDiff_RestMean_LOG
count	2065.000000
mean	inf
std	NaN
min	2.079442
25%	3.092859
50%	3.494100
75%	4.168703
max	inf

Threshold of FoldDiff: 16

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	8.960000e+02	896.000000	896.000000	896.000000
mean	2.031709e+04	389.906250	201.989397	127.457483
std	1.092277e+05	1657.088913	1000.941265	579.578809
min	1.600000e+01	0.000000	0.000000	16.000000

25%	2.440000e+02	6.000000	2.750000	21.266304
50%	1.468000e+03	28.500000	13.250000	30.992188
75%	8.493000e+03	143.250000	67.312500	68.844626
max	2.253663e+06	31253.000000	21759.250000	12412.000000

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG	\
count	896.000000	896.000000	896.000000	896.000000	
mean	inf	7.345930	3.801458	-inf	
std	NaN	2.296912	1.033031	NaN	
min	19.750520	2.772589	2.772589	-inf	
25%	47.694318	5.497168	3.057124	1.011601	
50%	74.615385	7.291656	3.433735	2.583998	
75%	172.000000	9.046993	4.231837	4.209345	
max	inf	14.628067	9.426419	9.987794	

	FoldDiff_RestMean_LOG
count	896.000000
mean	inf
std	NaN
min	2.983180
25%	3.864812
50%	4.312346
75%	5.147494
max	inf

Threshold of FoldDiff: 32

	LS_EXP	SecondMax	RestMean	FoldDiff	\
count	4.350000e+02	435.000000	435.000000	435.000000	
mean	3.073044e+04	296.452874	151.092529	238.972031	
std	1.519043e+05	1758.203487	1122.729190	817.602619	
min	3.300000e+01	0.000000	0.000000	32.125000	
25%	6.830000e+02	6.500000	3.000000	44.006400	
50%	3.270000e+03	31.000000	14.250000	74.452830	
75%	1.319700e+04	123.500000	55.875000	161.126961	
max	2.253663e+06	31253.000000	21759.250000	12412.000000	

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG	\
count	435.000000	435.000000	435.000000	435.000000	
mean	inf	8.016322	4.564515	-inf	
std	NaN	2.146121	1.011964	NaN	
min	42.947702	3.496508	3.469635	-inf	
25%	104.032356	6.526456	3.784335	1.098612	
50%	175.913043	8.092545	4.310166	2.656757	
75%	359.627530	9.487694	5.082192	4.023095	
max	inf	14.628067	9.426419	9.987794	

	FoldDiff_RestMean_LOG
count	435.000000
mean	inf

```

std          NaN
min          3.759983
25%          4.644702
50%          5.169990
75%          5.885068
max          inf

```

Threshold of FoldDiff: 64

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	2.430000e+02	243.000000	243.000000	243.000000
mean	4.192780e+04	184.621399	80.042181	392.949462
std	1.882614e+05	686.621839	282.602225	1069.972609
min	7.500000e+01	0.000000	0.000000	64.000000
25%	1.528500e+03	7.000000	3.500000	90.302020
50%	5.698000e+03	32.000000	16.250000	135.000000
75%	1.771950e+04	121.000000	52.875000	275.230000
max	2.253663e+06	6635.000000	2439.000000	12412.000000

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG \
count	243.000000	243.000000	243.000000	243.000000
mean	inf	8.639722	5.192860	-inf
std	NaN	1.927644	0.954800	NaN
min	79.512195	4.317488	4.158883	-inf
25%	198.666667	7.331232	4.503147	1.252763
50%	329.444444	8.647871	4.905275	2.788093
75%	640.277551	9.782410	5.617553	3.967794
max	inf	14.628067	9.426419	7.799343

```

FoldDiff_RestMean_LOG
count          243.000000
mean          inf
std          NaN
min          4.375910
25%          5.291606
50%          5.797408
75%          6.461902
max          inf

```

Threshold of FoldDiff: 128

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	1.280000e+02	128.000000	128.000000	128.000000
mean	6.679294e+04	198.757812	84.787109	665.200762
std	2.526545e+05	719.674442	312.341413	1422.451233
min	1.440000e+02	0.000000	0.000000	128.200000
25%	2.808000e+03	6.750000	3.250000	190.966121
50%	9.453000e+03	26.500000	13.625000	263.228205
75%	3.939975e+04	121.250000	51.312500	405.291451
max	2.253663e+06	5944.000000	2439.000000	12412.000000

```

FoldDiff_RestMean LS_EXP_LOG FoldDiff_LOG RestMean_LOG \

```

count	128.000000	128.000000	128.000000	128.000000
mean	inf	9.205471	5.833815	-inf
std	NaN	1.893452	0.909382	NaN
min	187.359116	4.969813	4.853592	-inf
25%	380.069594	7.940212	5.252095	1.178655
50%	611.012642	9.153855	5.573018	2.611527
75%	1152.523682	10.581488	6.004585	3.937566
max	inf	14.628067	9.426419	7.799343

#### FoldDiff\_RestMean\_LOG

count	128.000000
mean	inf
std	NaN
min	5.233027
25%	5.940353
50%	6.415118
75%	7.049708
max	inf

Threshold of FoldDiff: 256

	LS_EXP	SecondMax	RestMean	FoldDiff \
count	6.800000e+01	68.000000	68.000000	68.000000
mean	1.041002e+05	262.176471	116.525735	1089.632088
std	3.396591e+05	948.722172	418.548195	1855.841604
min	4.350000e+02	0.000000	0.000000	256.062500
25%	3.520250e+03	5.750000	2.500000	321.370004
50%	1.222350e+04	21.000000	8.500000	402.690848
75%	6.791150e+04	141.250000	59.562500	894.000000
max	2.253663e+06	5944.000000	2439.000000	12412.000000

	FoldDiff_RestMean	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG \
count	68.000000	68.000000	68.000000	68.000000
mean	inf	9.667749	6.395987	-inf
std	NaN	1.882576	0.921161	NaN
min	379.029821	6.075346	5.545422	-inf
25%	695.024316	8.166179	5.772592	0.916291
50%	1074.166667	9.410997	5.998169	2.140066
75%	1910.908824	11.125859	6.795521	4.086894
max	inf	14.628067	9.426419	7.799343

#### FoldDiff\_RestMean\_LOG

count	68.000000
mean	inf
std	NaN
min	5.937615
25%	6.543823
50%	6.979229
75%	7.553861
max	inf

Threshold of FoldDiff: 512

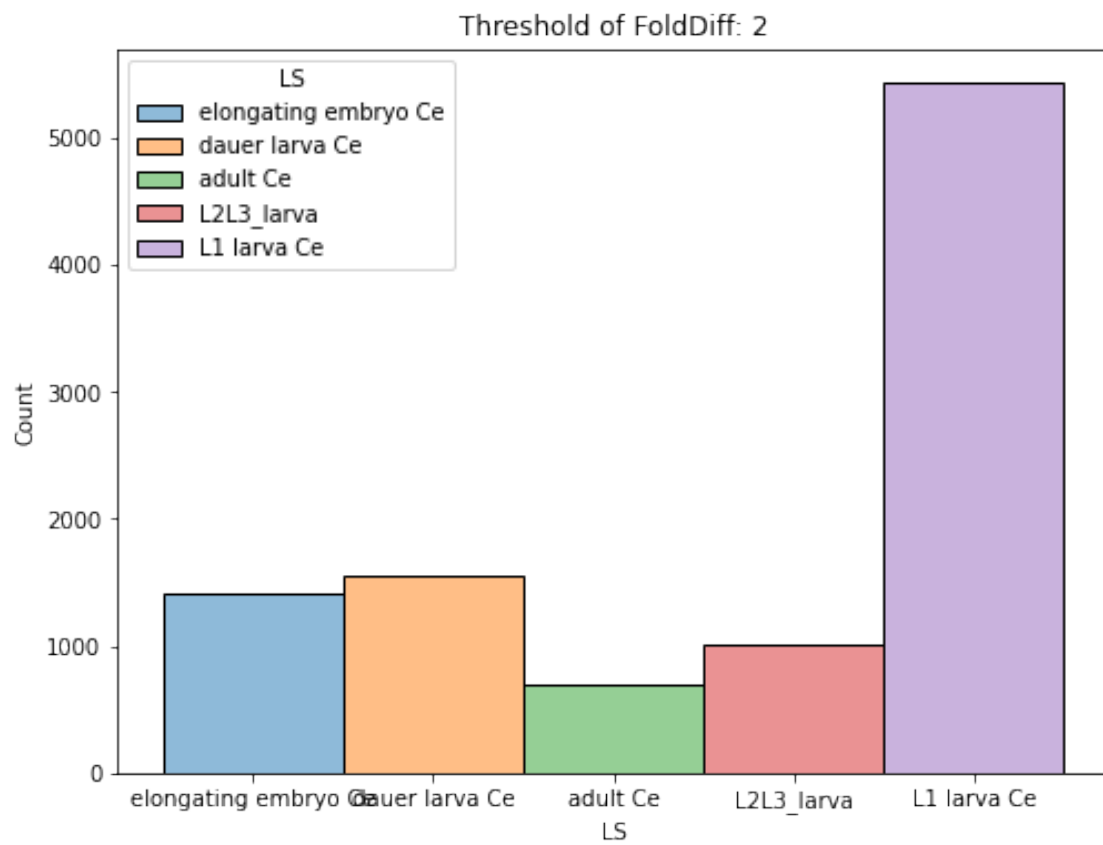
	LS_EXP	SecondMax	RestMean	FoldDiff	FoldDiff_RestMean	\
count	25.00000	25.000000	25.000000	25.000000	25.000000	
mean	39358.52000	16.680000	6.890000	2371.482954	inf	
std	66170.68322	25.450475	9.392783	2626.937144	NaN	
min	677.00000	0.000000	0.000000	540.750000	1236.000000	
25%	3550.00000	3.000000	1.250000	826.666667	1724.666667	
50%	11140.00000	6.000000	3.250000	1293.500000	4030.666667	
75%	18451.00000	19.000000	8.000000	2785.000000	8912.000000	
max	235883.00000	95.000000	33.250000	12412.000000	inf	

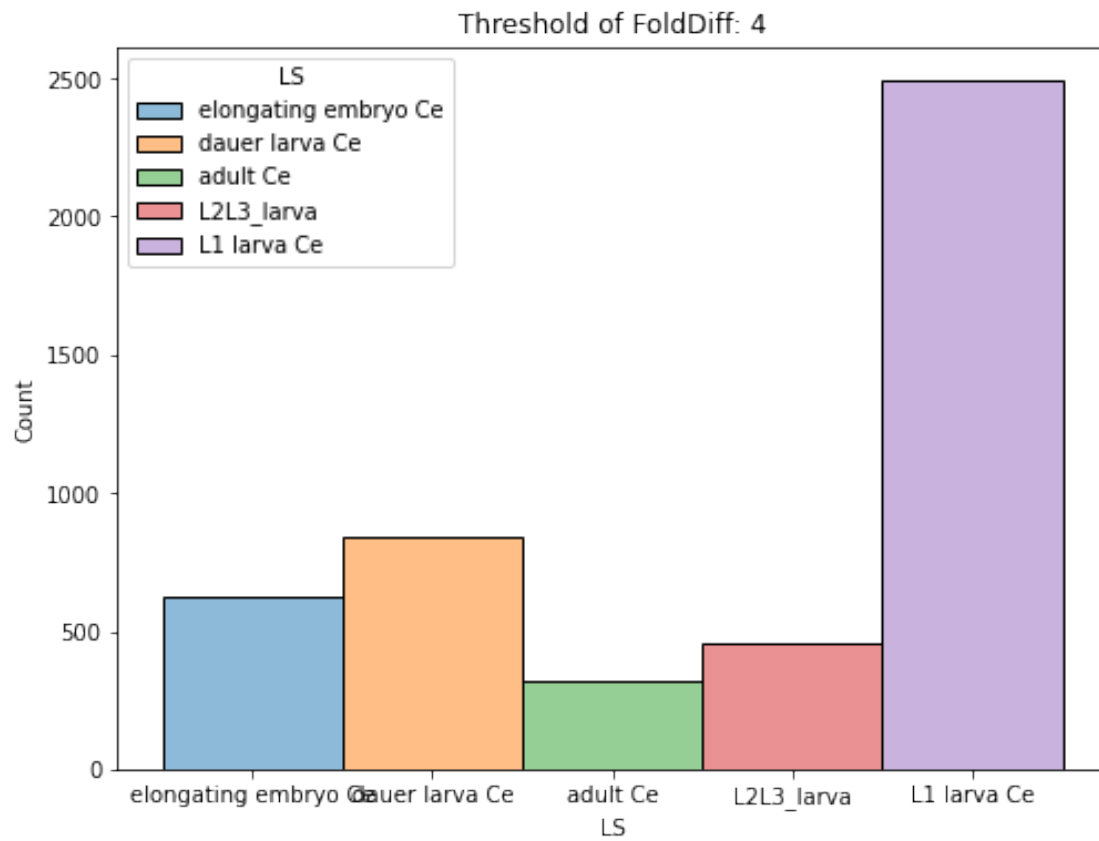
	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG	FoldDiff_RestMean_LOG	
count	25.000000	25.000000	25.000000	25.000000	
mean	9.341956	7.378386	-inf	inf	
std	1.612699	0.848040	NaN	NaN	
min	6.517671	6.292957	-inf	7.119636	
25%	8.174703	6.717402	0.223144	7.452789	
50%	9.318298	7.165107	1.178655	8.301687	
75%	9.822874	7.932003	2.079442	9.095154	
max	12.371091	9.426419	3.504055	inf	

Threshold of FoldDiff: 1024

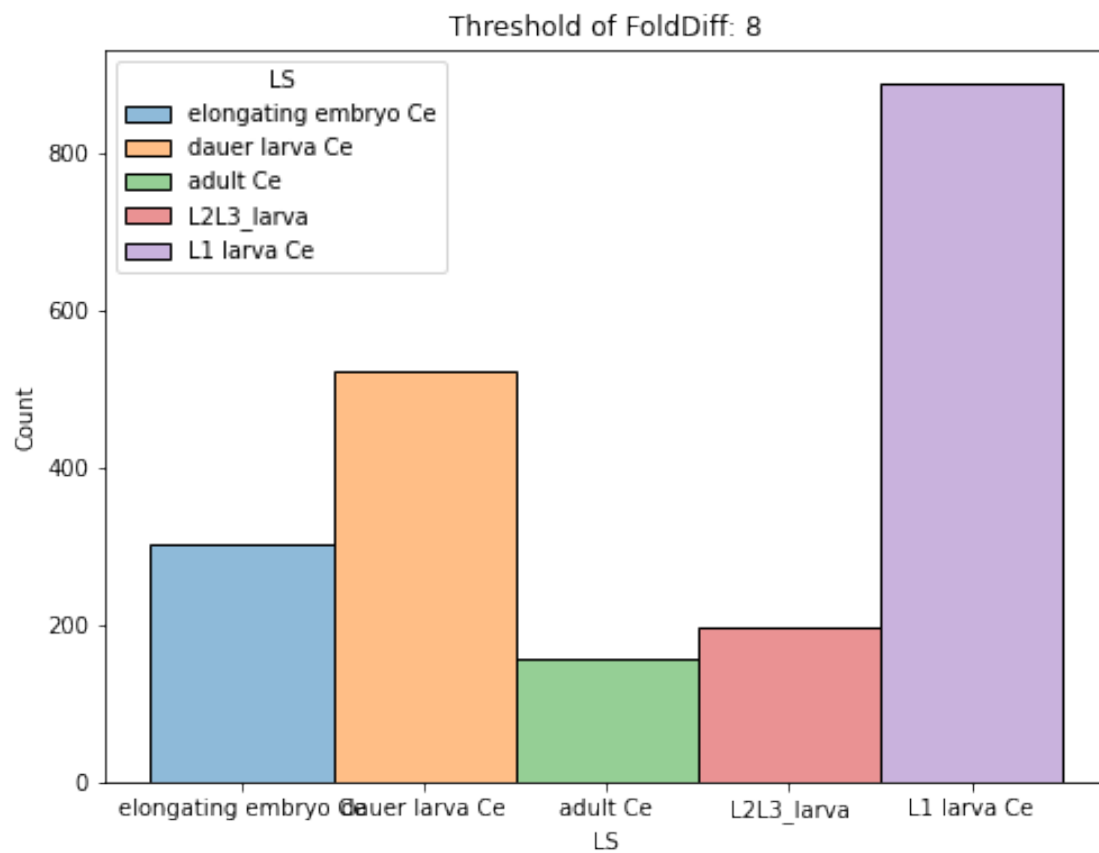
	LS_EXP	SecondMax	RestMean	FoldDiff	FoldDiff_RestMean	\
count	15.000000	15.000000	15.000000	15.000000	15.000000	
mean	61883.400000	22.866667	9.433333	3457.096160	9341.260444	
std	78268.637283	31.217822	11.307972	2942.840547	11710.916859	
min	3023.000000	1.000000	0.250000	1130.571429	1663.866667	
25%	11010.500000	4.000000	1.375000	1588.897368	4395.754386	
50%	15522.000000	10.000000	4.250000	2720.250000	5677.230769	
75%	127015.000000	23.500000	10.750000	4187.228649	9654.957895	
max	235883.000000	95.000000	33.250000	12412.000000	49648.000000	

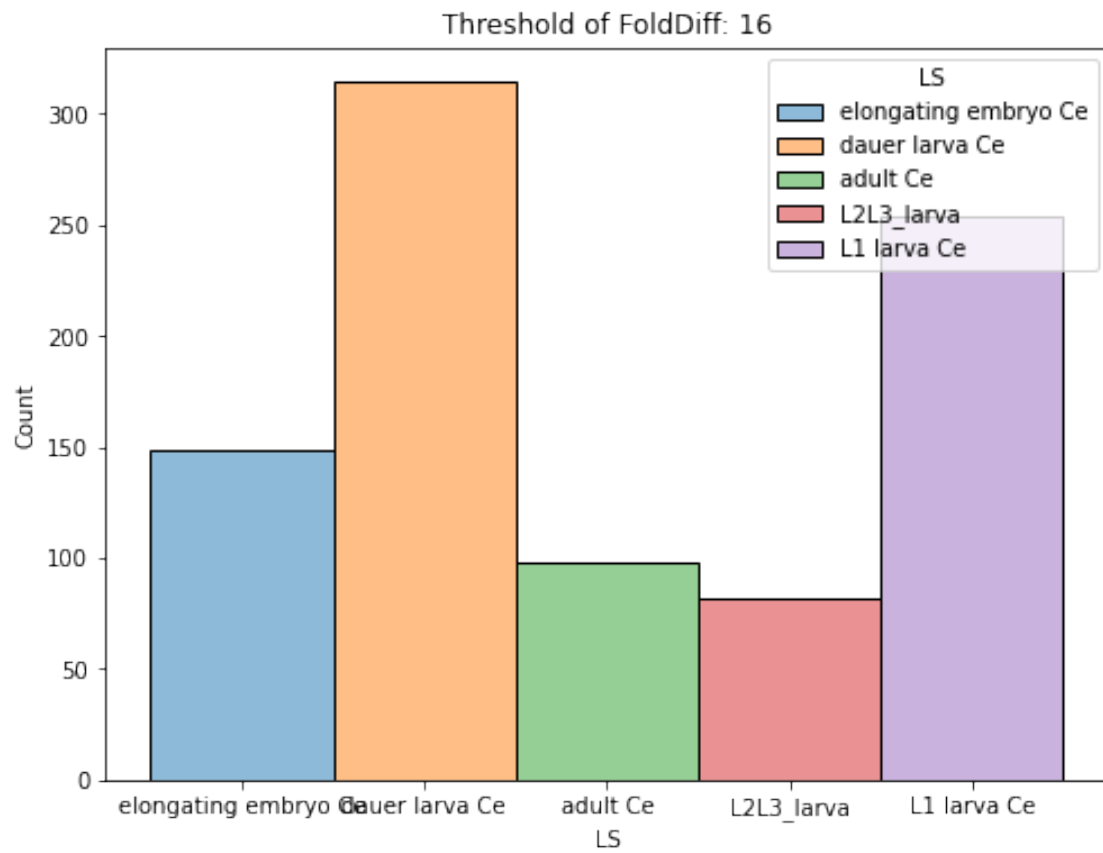
	LS_EXP_LOG	FoldDiff_LOG	RestMean_LOG	FoldDiff_RestMean_LOG	
count	15.000000	15.000000	15.000000	15.000000	
mean	10.139472	7.901033	1.405217	8.734256	
std	1.425323	0.691138	1.512399	0.871807	
min	8.014005	7.030478	-1.386294	7.416899	
25%	9.306535	7.369608	0.314304	8.384933	
50%	9.650014	7.908479	1.446919	8.644219	
75%	11.720771	8.339777	2.361477	9.174815	
max	12.371091	9.426419	3.504055	10.812713	

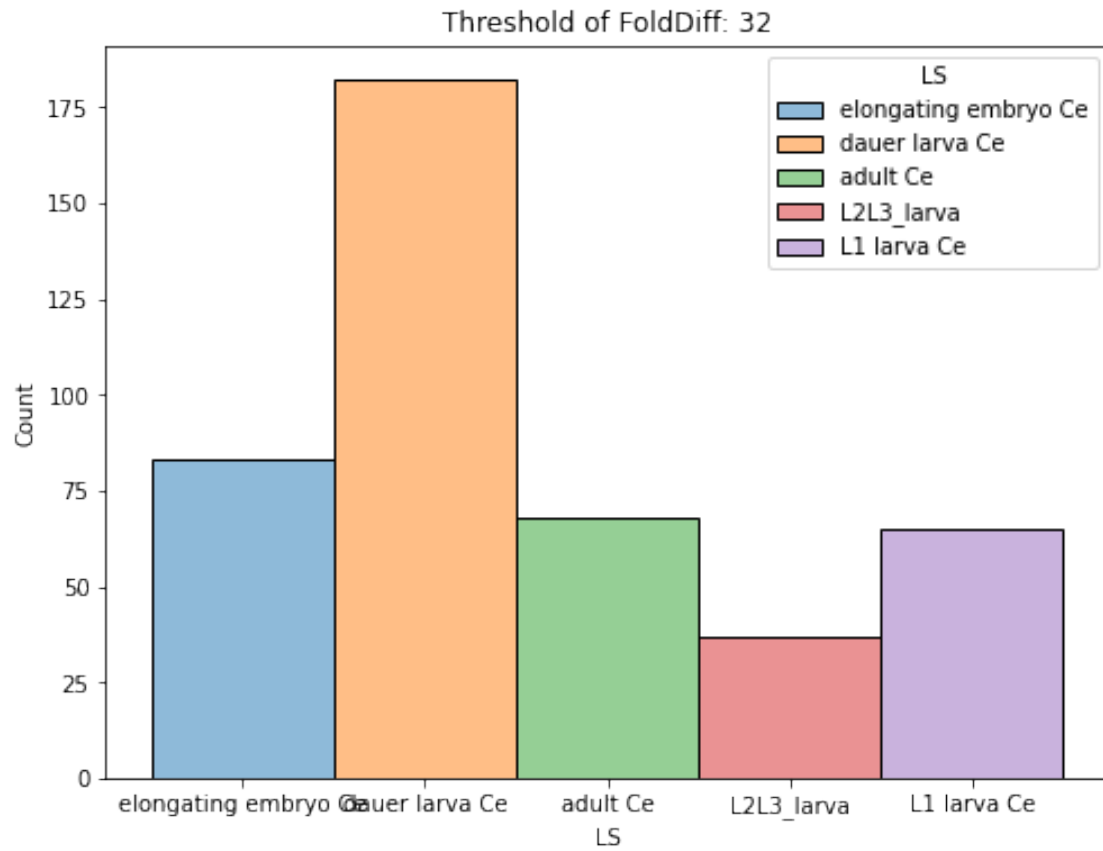


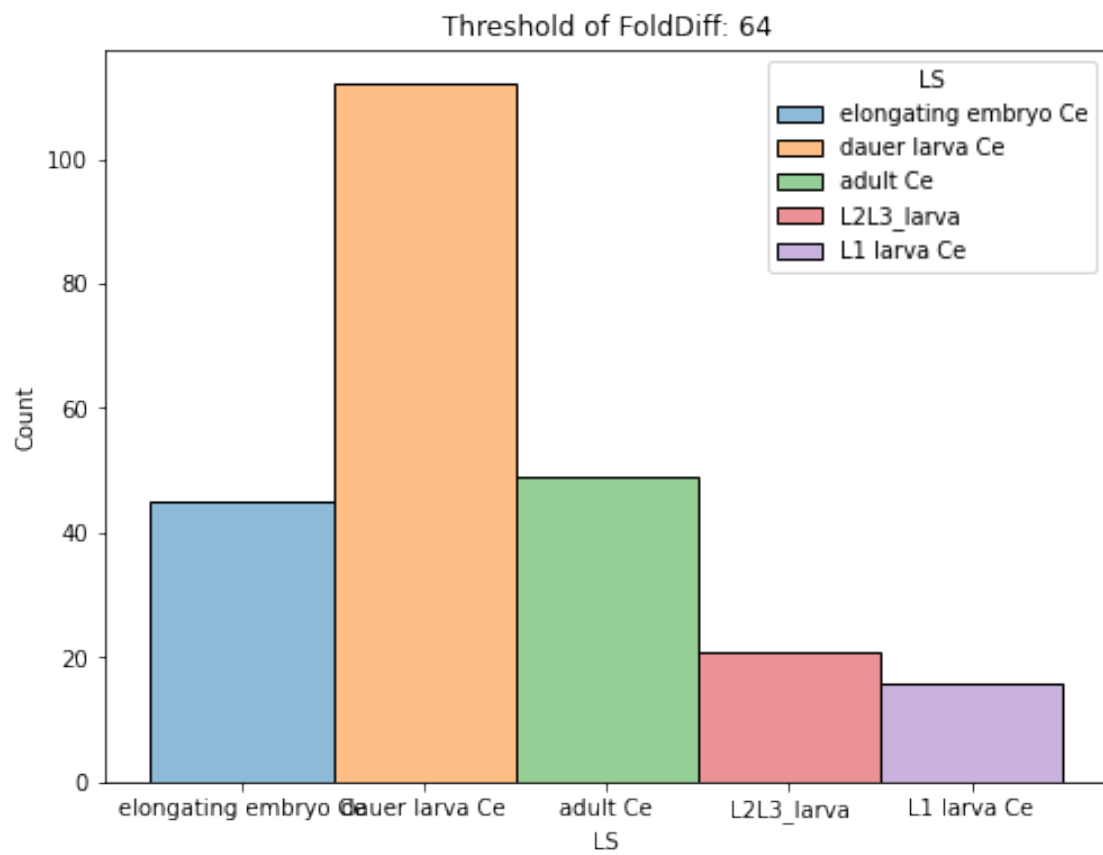


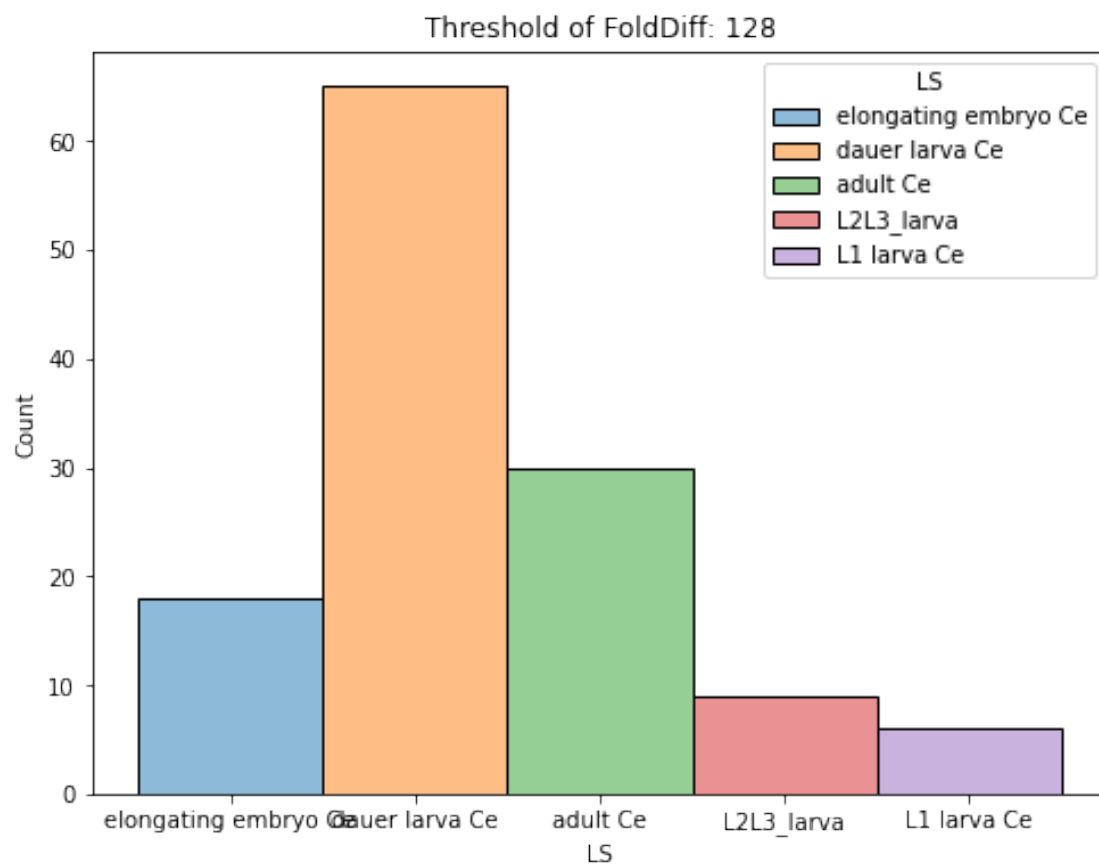


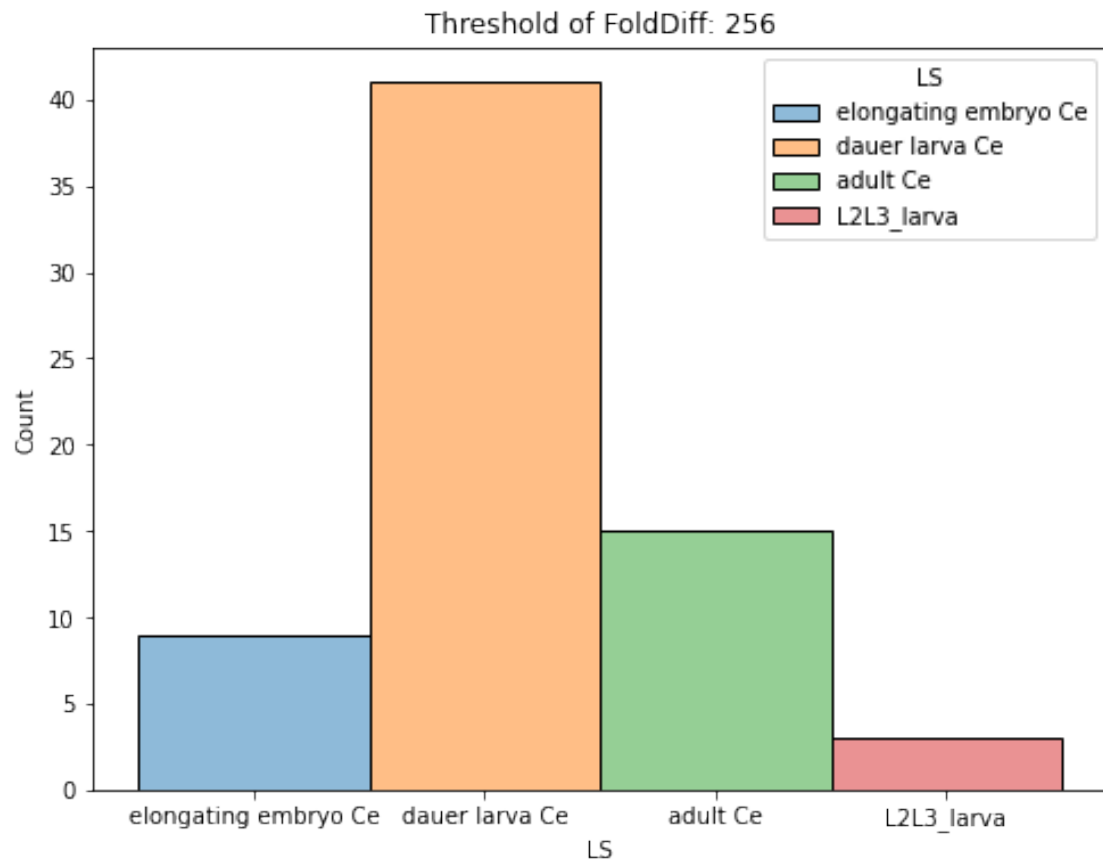


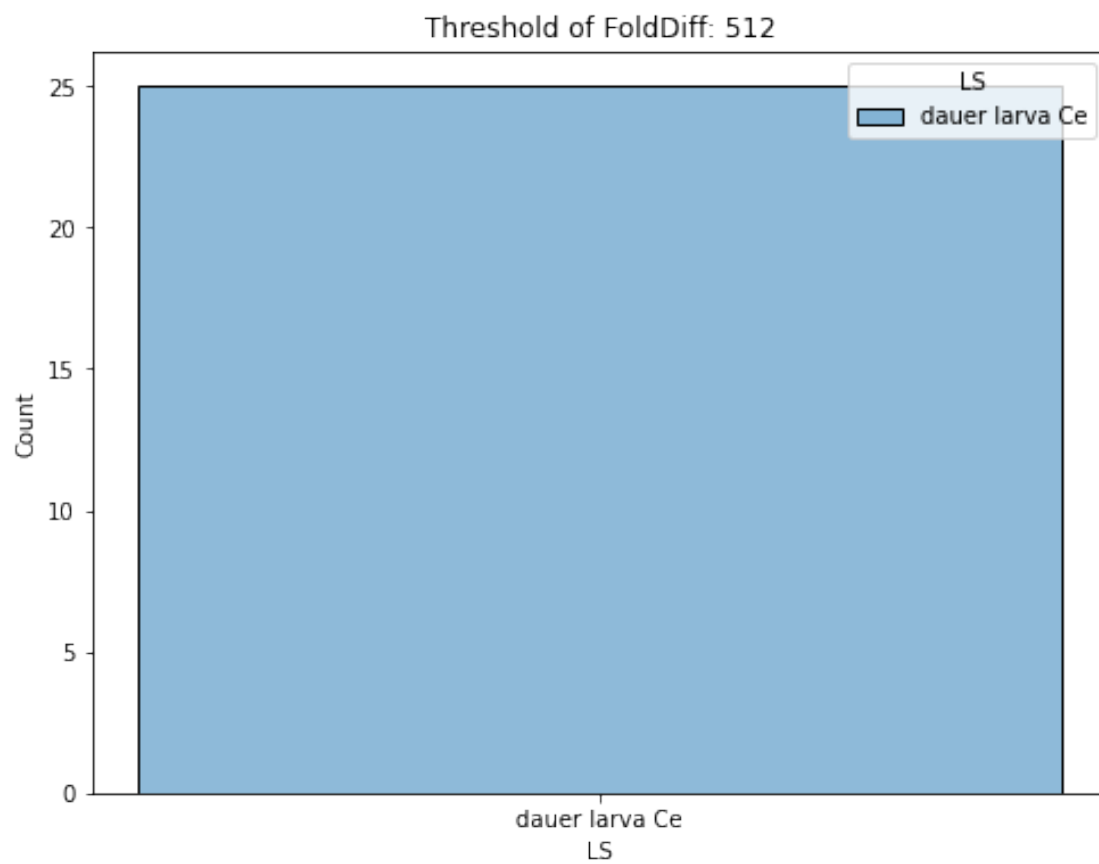


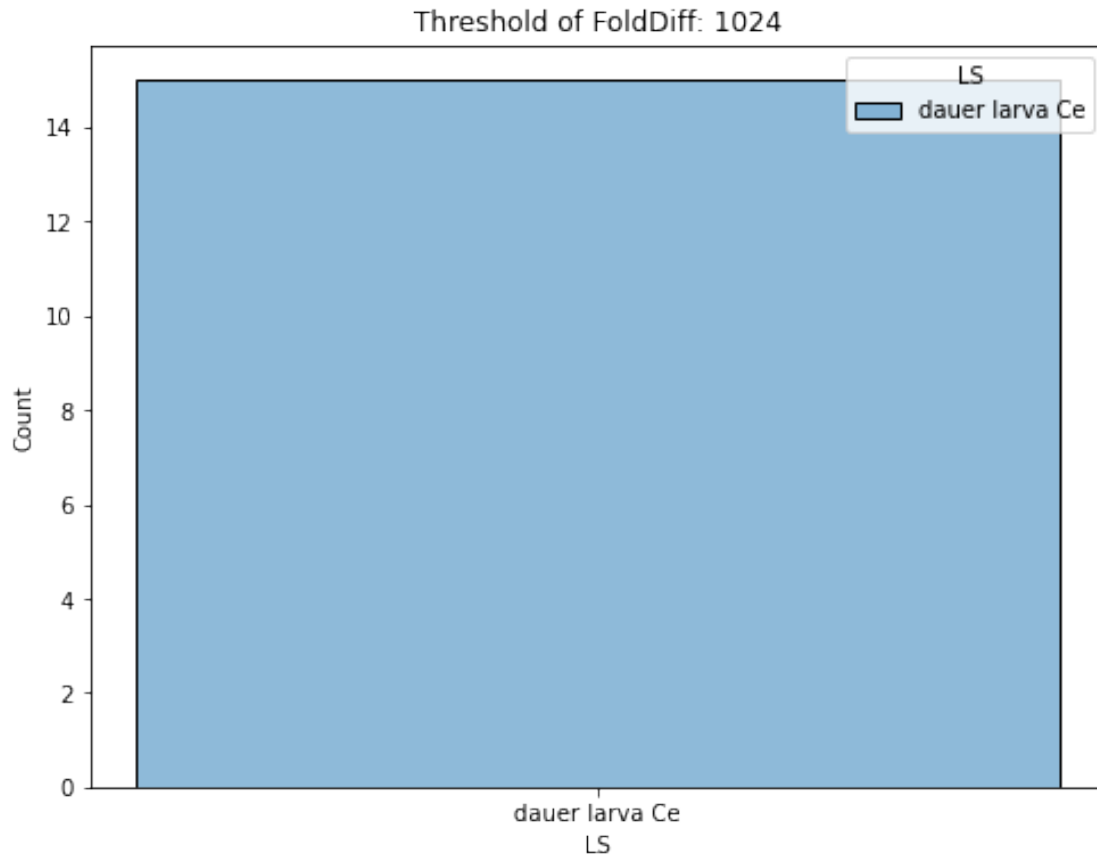










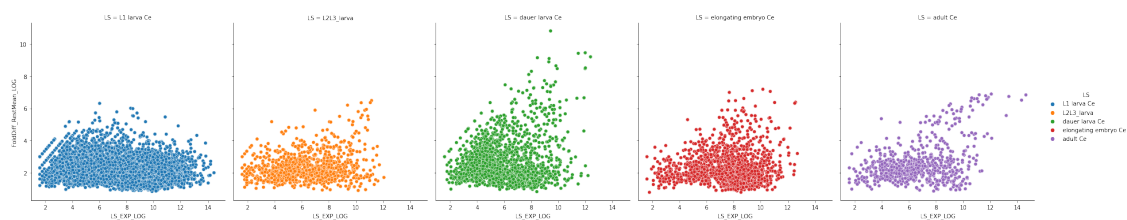


Below is the same figures using foldDiff\_RestMean rather than FoldDiff(second max)

```
[20]: ax3=plt.figure(figsize=[20,20])
sns.relplot(
    data=ls_data,x='LS_EXP_LOG', y="FoldDiff_RestMean_LOG",
    col="LS", hue="LS",
    kind="scatter"
)
```

[20]: <seaborn.axisgrid.FacetGrid at 0x7f67f807d820>

<Figure size 1440x1440 with 0 Axes>

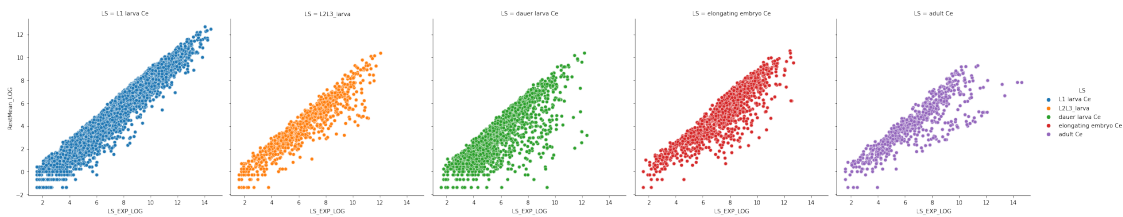




```
[11]: ax4=plt.figure(figsize=[20,20])

sns.relplot(
    data=ls_data,x='LS_EXP_LOG', y="RestMean_LOG",
    col="LS", hue="LS",
    kind="scatter"
)
```

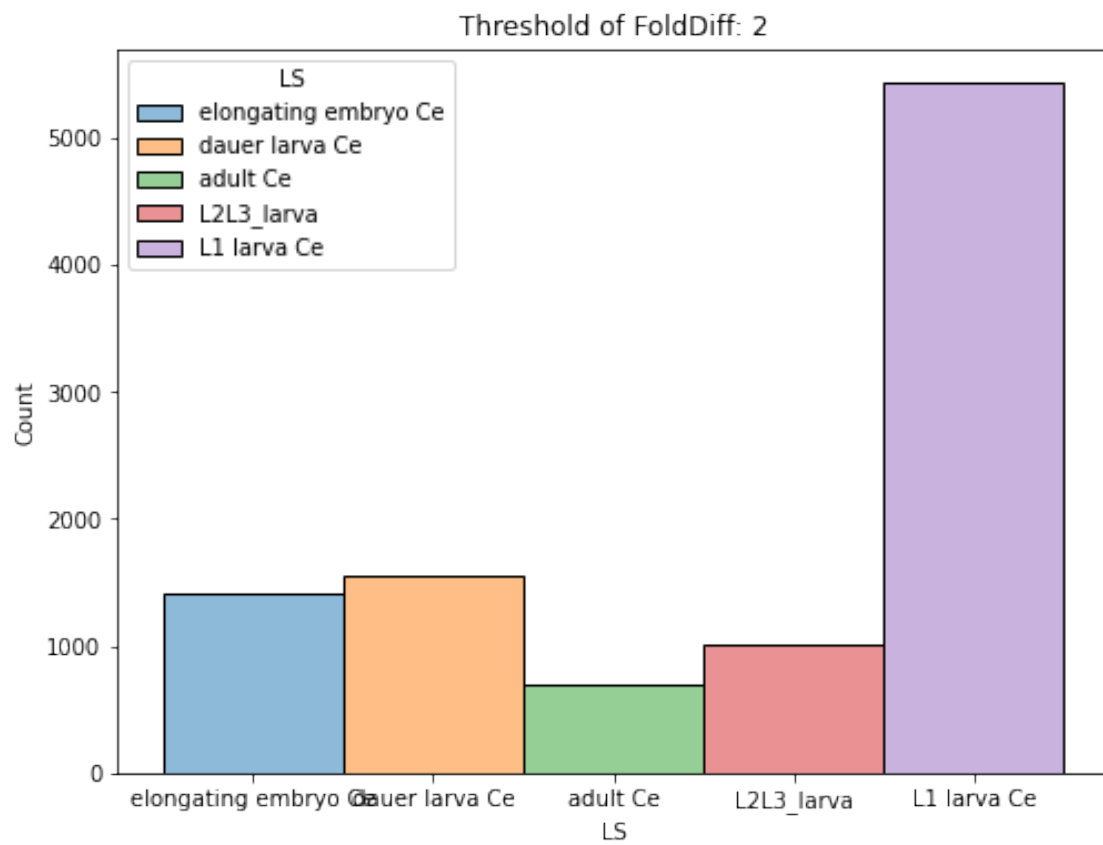
<Figure size 1440x1440 with 0 Axes>

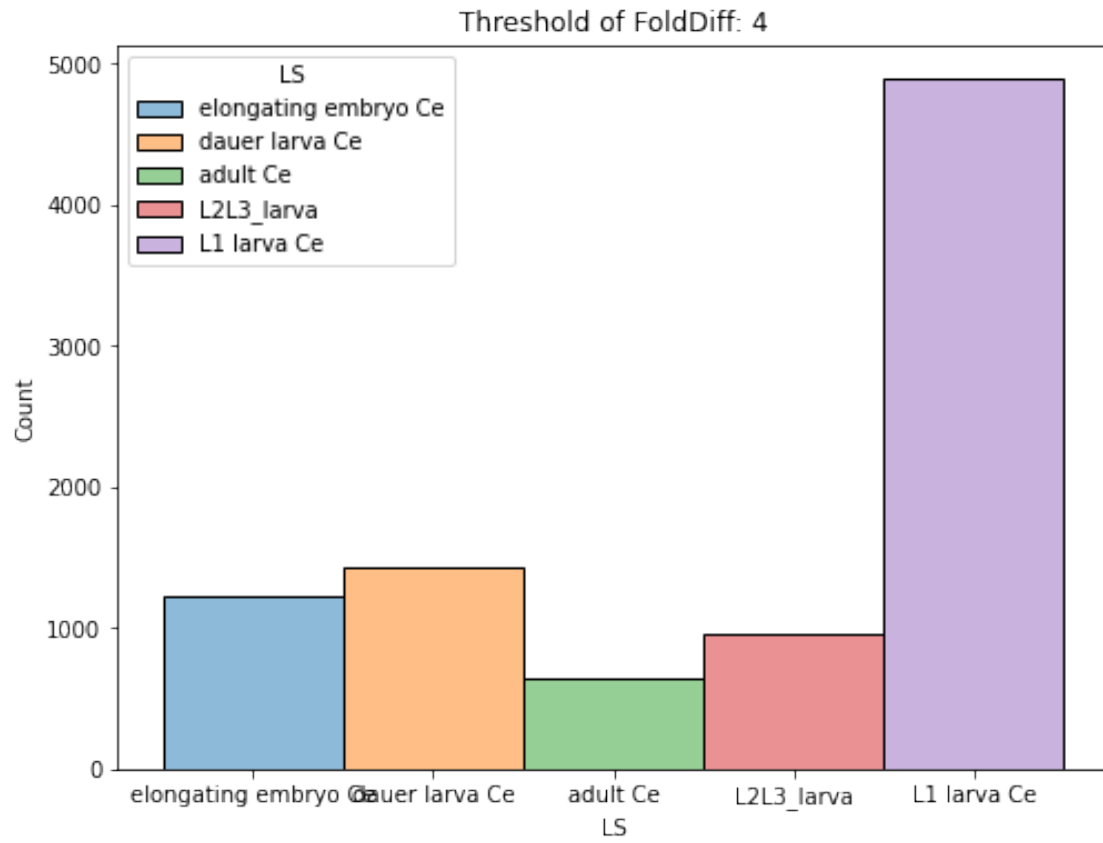


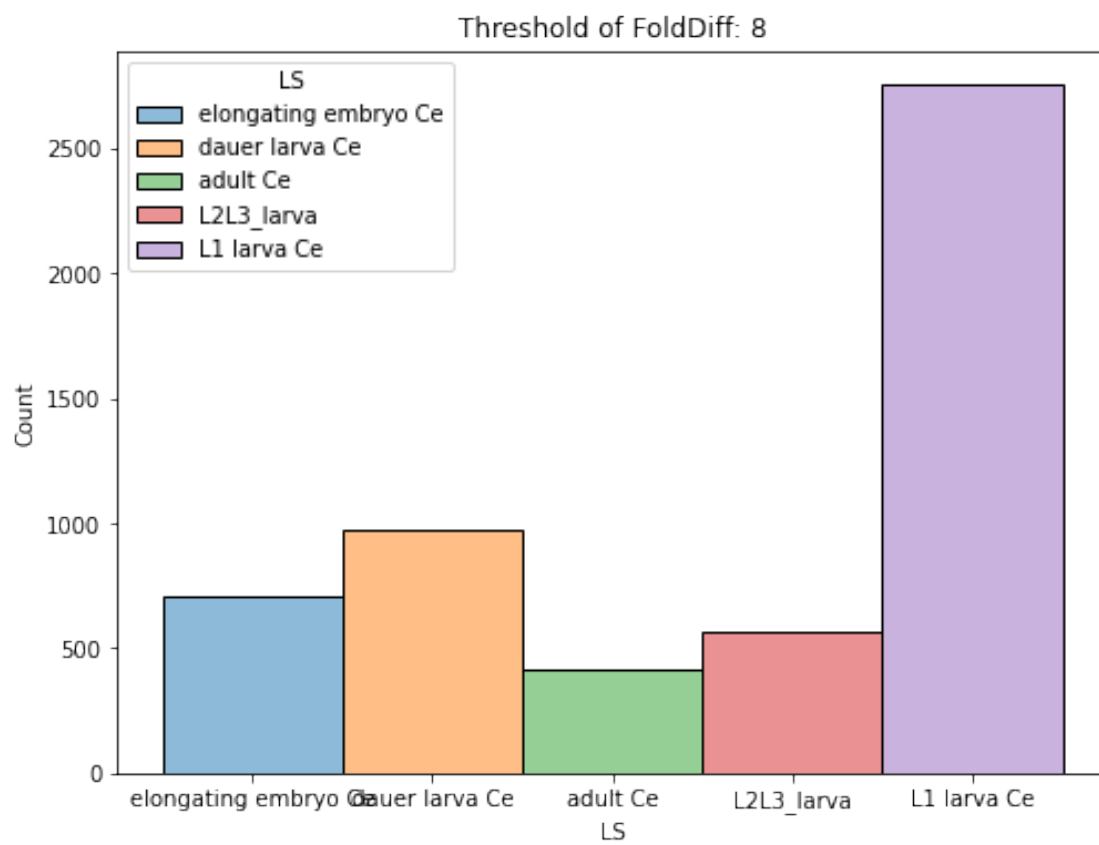
```
[12]: sorted_ls_data=ls_data.sort_values(['LS','FoldDiff_RestMean'],ascending=False)

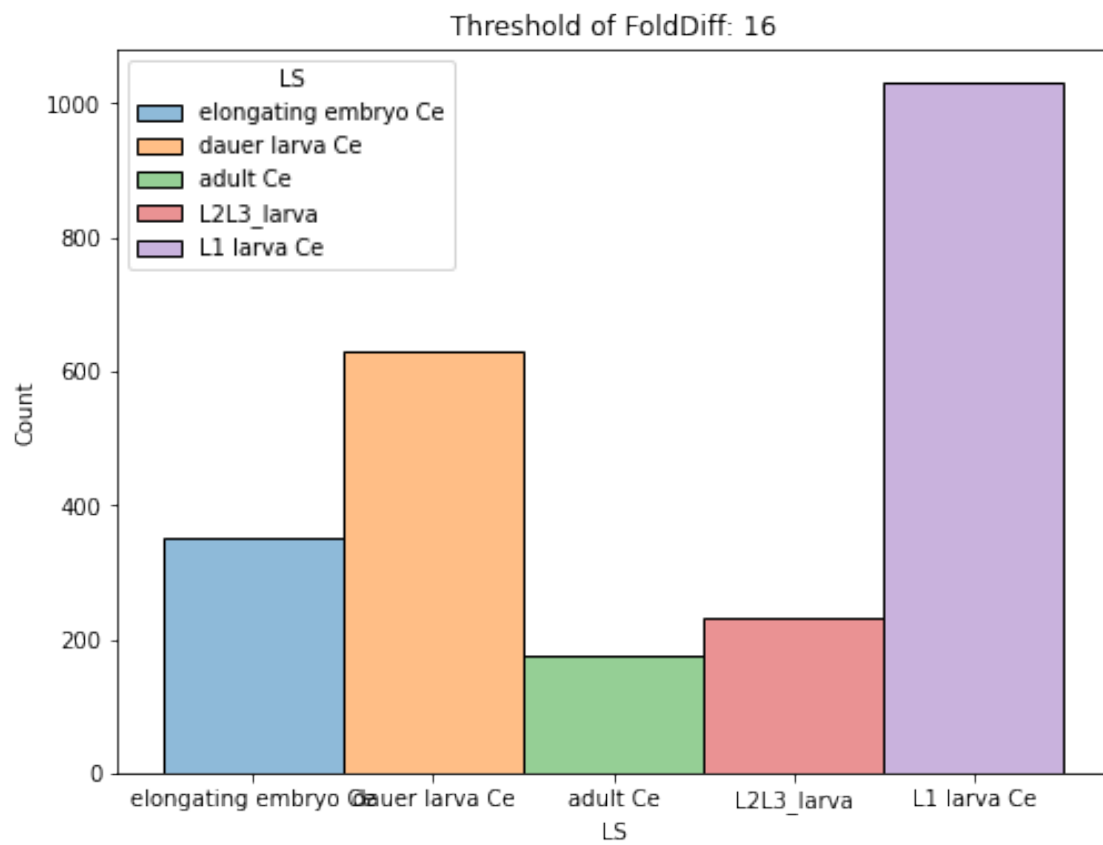
thresholds=[2**i for i in range(1,11)]

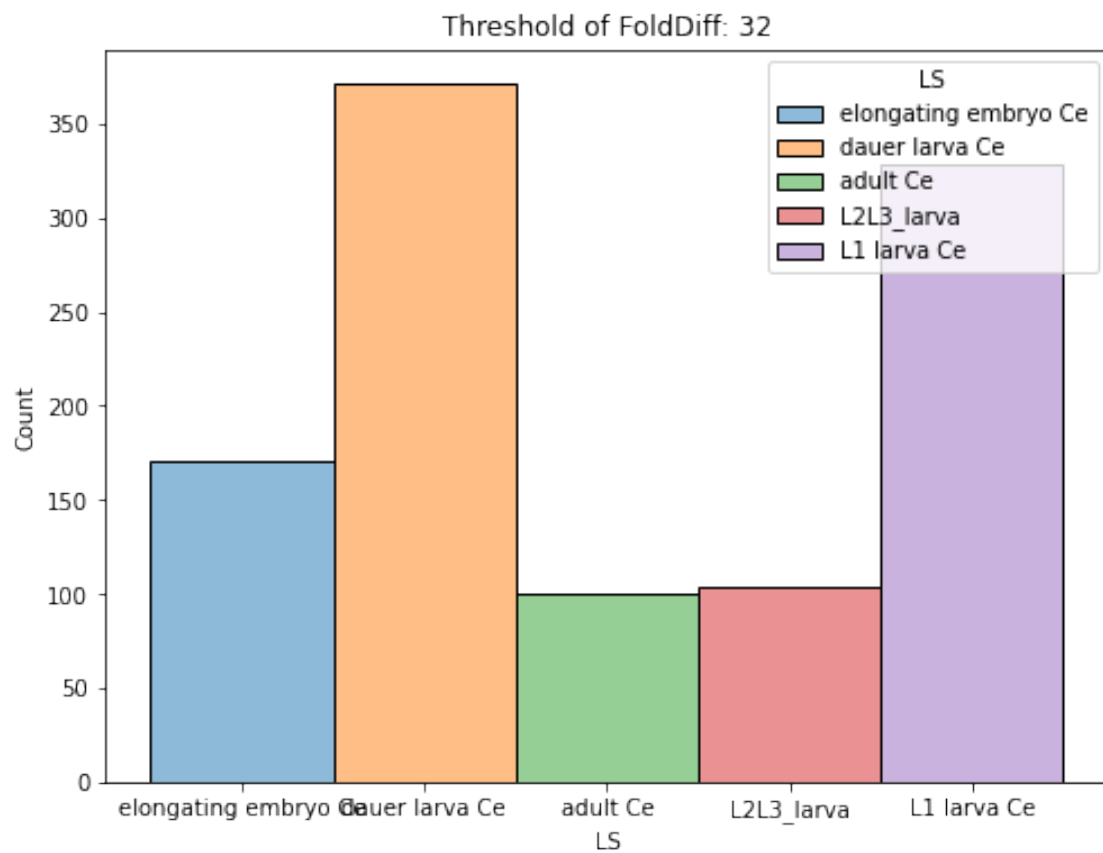
for threshold in thresholds:
    df_filtered=sorted_ls_data.loc[sorted_ls_data['FoldDiff_RestMean'] >=
    ↪threshold]
    df_count=df_filtered.groupby("LS").count()
    ax=plt.figure(figsize=[8,6])
    text=("Threshold of FoldDiff: "+ str(threshold))
    sns.histplot(df_filtered, x="LS",hue="LS").set_title(text)
```

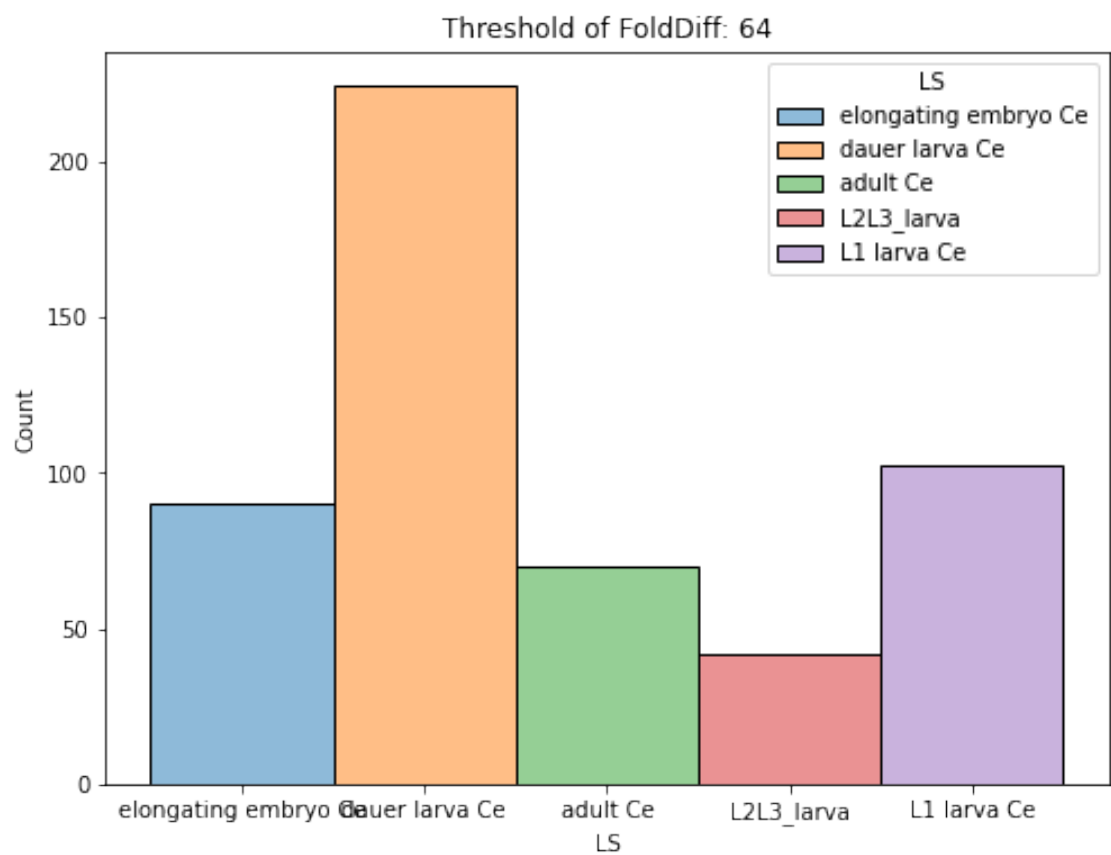


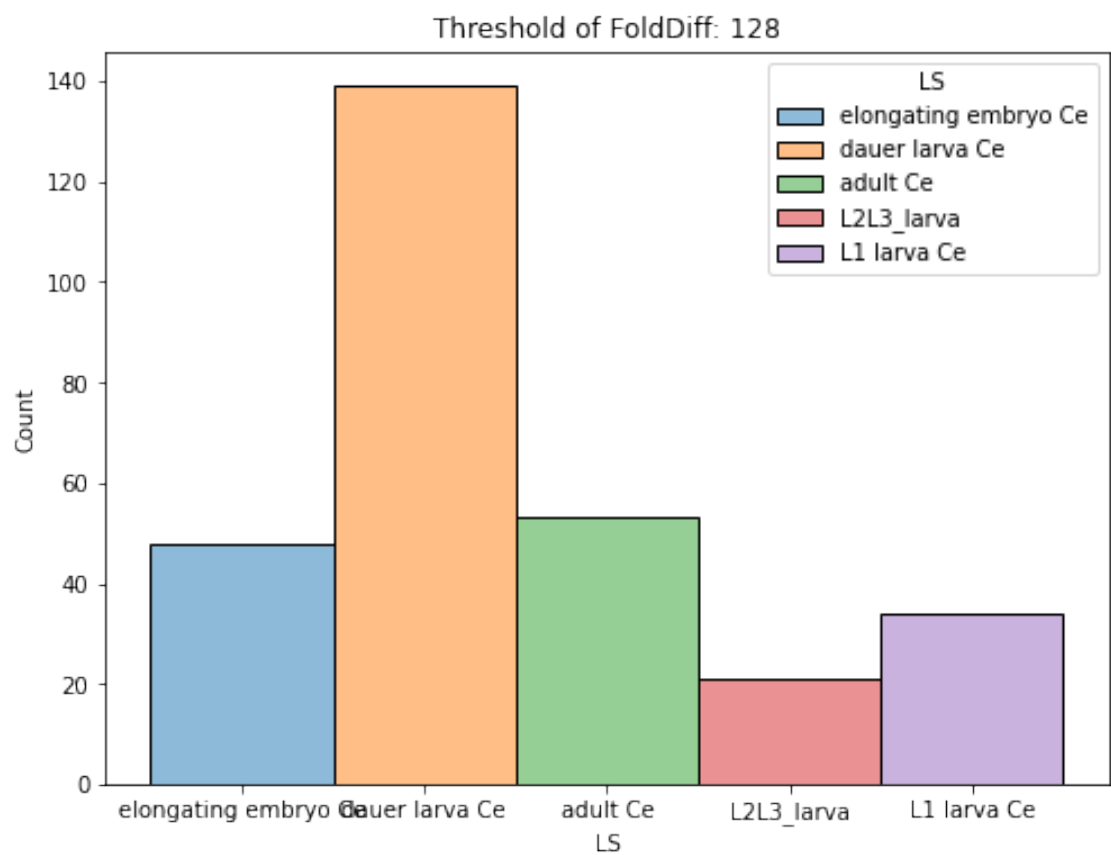




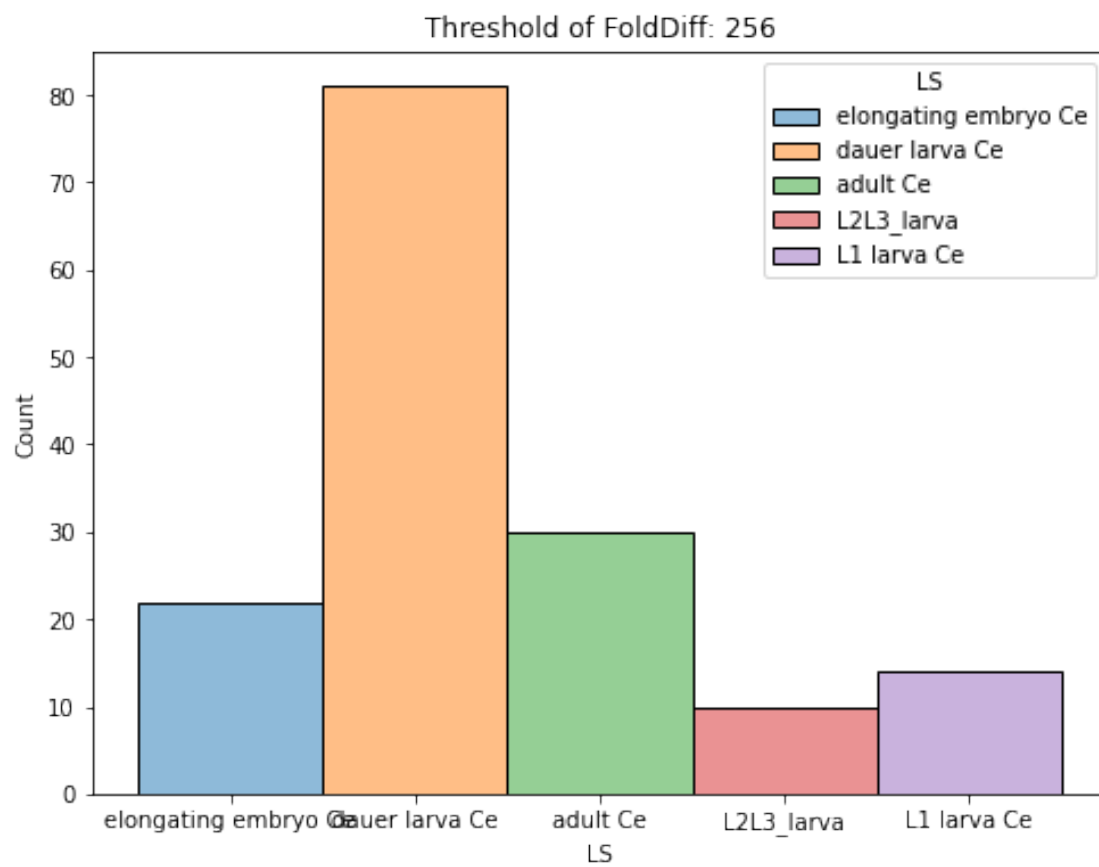


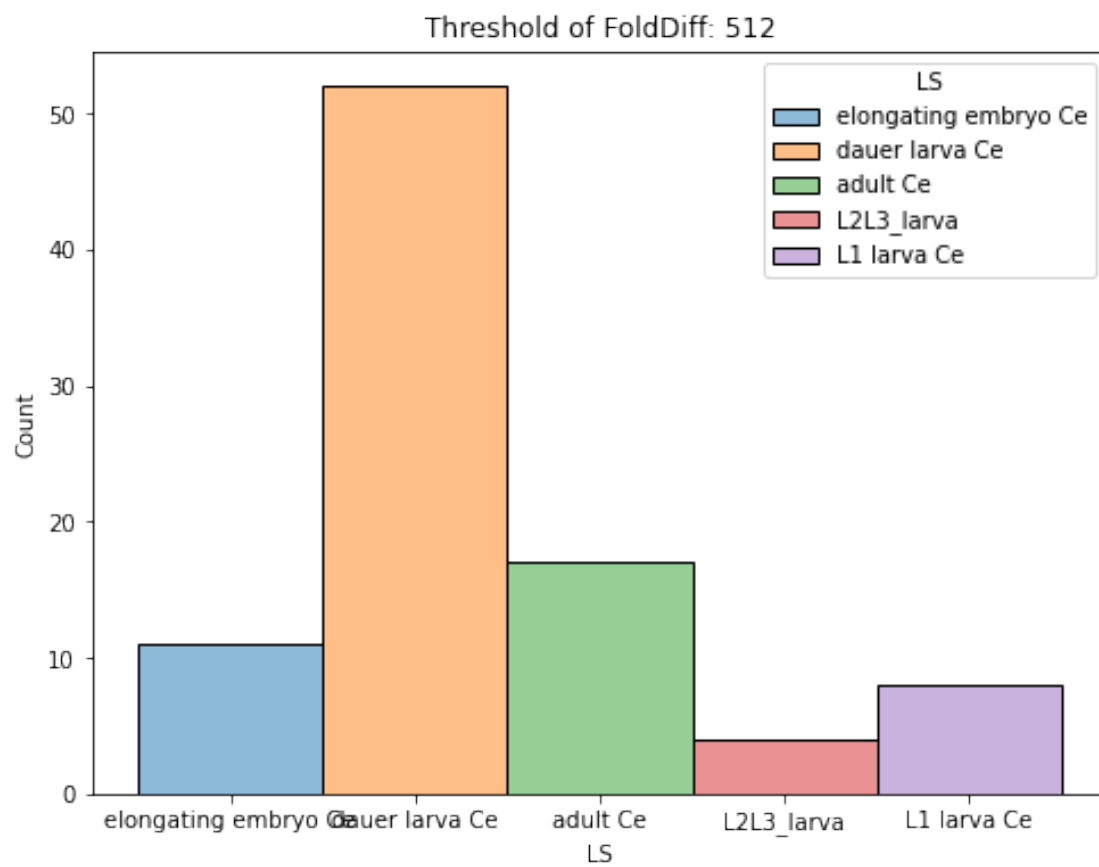


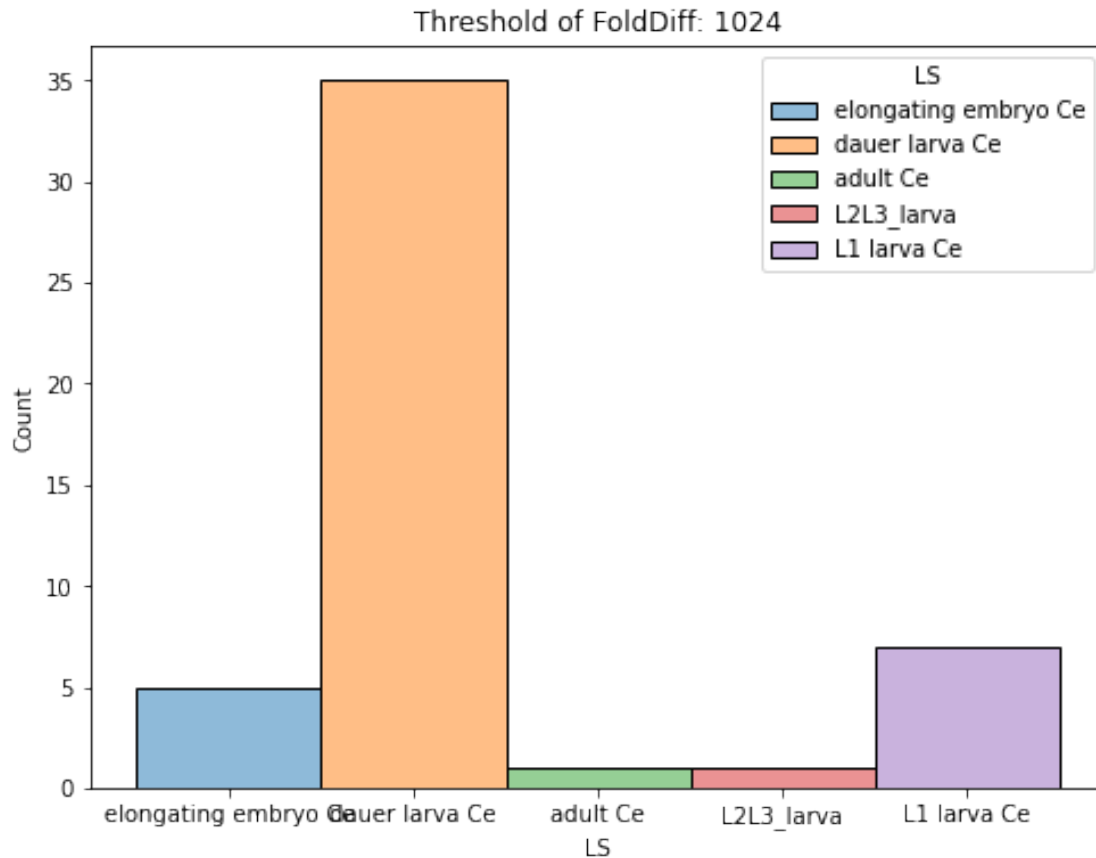










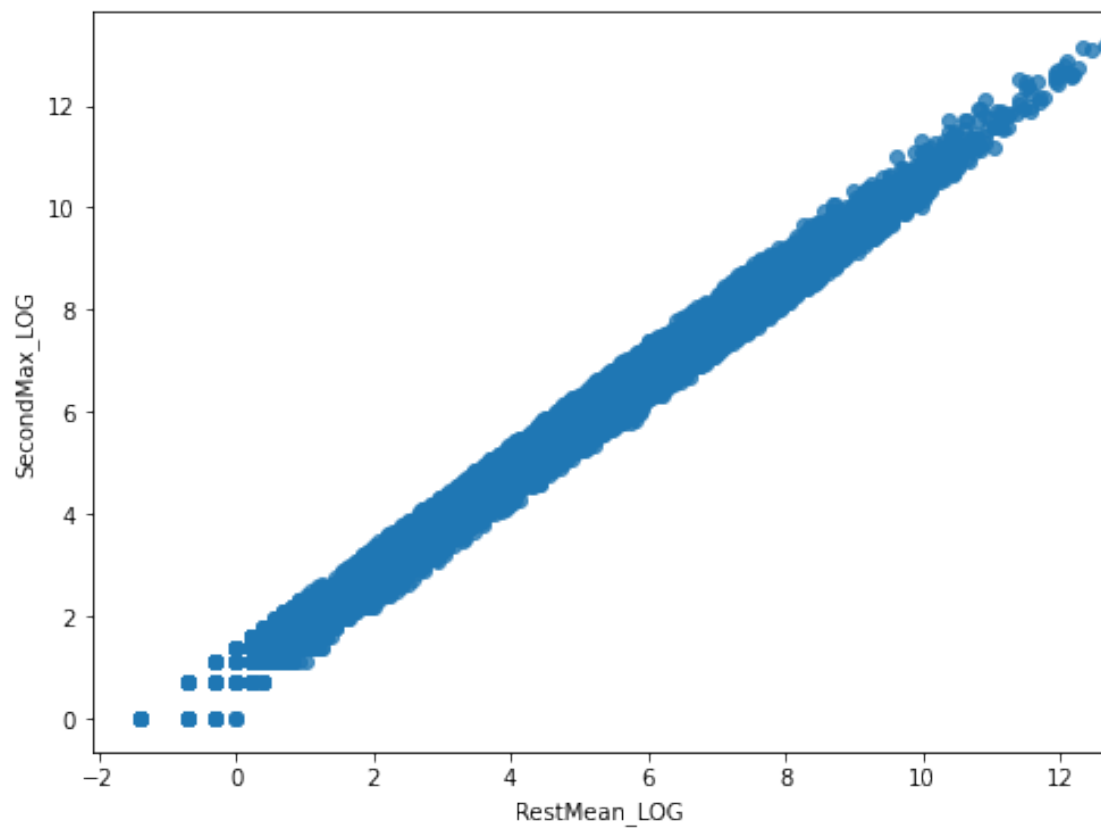


```
[28]: ax=plt.figure(figsize=[8,6])

ls_data["RestMean_LOG"]=np.log(ls_data["RestMean"])
ls_data["SecondMax_LOG"]=np.log(ls_data["SecondMax"])
sns.regplot(data=ls_data,x="RestMean_LOG",y="SecondMax_LOG")
```

```
/home/lu/.local/lib/python3.8/site-packages/numpy/core/function_base.py:144:
RuntimeWarning: invalid value encountered in multiply
    y *= step
/home/lu/.local/lib/python3.8/site-packages/numpy/core/function_base.py:154:
RuntimeWarning: invalid value encountered in add
    y += start
```

```
[28]: <AxesSubplot:xlabel='RestMean_LOG', ylabel='SecondMax_LOG'>
```



[ ]: