

# v19\_220601\_zerofilled

June 20, 2022

## 0.1 Load Data

```
[1]: import pandas as pd
import numpy as np
import os
import glob
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
import matplotlib.pyplot as plt
import missingno as msno
from sklearn.cluster import KMeans
import seaborn as sns
from matplotlib_venn import venn3
from scipy import stats
from sklearn.metrics import silhouette_score
%matplotlib inline
import datetime
from chord import Chord
```

```
[2]: date = str(datetime.date.today())
print(date)
```

2022-06-08

```
[3]: #set working directory
os.chdir('/Users/kavyasharman/Documents/Projects/microLESA/')

```

```
[4]: #load data as df
pGroups = pd.read_csv('/Users/kavyasharman/Documents/Projects/microLESA/data/
→proteinGroups.txt', delimiter = '\t', low_memory = False)
pGroups.shape
```

```
[4]: (3760, 371)
```

```
[5]: pGroups.head()
```

```
[5]:
```

	Protein IDs \
0	G3UZW7;A0A023T778;Q9CQL1;P61327
1	O70589;A0A067XG53;F6Y9I5
2	A0A140T8M9;A0A140T8M0;A0A0B4J1I0;A0A075B5N0;A0...

```

3      AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1
4      AOA075B5P3;AOA0A6YVP0;P01867

```

```

Majority protein IDs Peptide counts (all) \
0      G3UZW7;AOA023T778;Q9CQL1;P61327      1;1;1;1
1      070589;AOA067XG53;F6Y9I5      2;2;1
2      AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AO...      1;1;1;1;1;1;1;1;1;1
3      AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1      2;1;1;1
4      AOA075B5P3;AOA0A6YVP0;P01867      3;3;3

```

```

Peptide counts (razor+unique) Peptide counts (unique) \
0      1;1;1;1      1;1;1;1
1      2;2;1      2;2;1
2      1;1;1;1;1;1;1;1;1      1;1;1;1;1;1;1;1;1
3      2;1;1;1      2;1;1;1
4      3;3;3      3;3;3

```

```

Protein names \
0      Protein mago nashi homolog 2;Protein mago nash...
1      Peripheral plasma membrane protein CASK
2      Ig kappa chain V-II region 26-10
3      NaN
4      Ig gamma-2B chain C region

```

```

Gene names \
0      Magohb;Magoh
1      Cask
2      Igkv1-110;Igkv1-35;Igkv1-99;Igkv1-115
3      Igkv5-39;Igkv5-43;Igkv5-45
4      Ighg2b;Igh-3

```

```

Fasta headers Number of proteins \
0      tr|G3UZW7|G3UZW7_MOUSE Protein mago nashi homo...      4
1      sp|070589|CSKP_MOUSE Peripheral plasma membran...      3
2      tr|AOA140T8M9|AOA140T8M9_MOUSE Immunoglobulin ...      9
3      tr|AOA075B5M7|AOA075B5M7_MOUSE Immunoglobulin ...      4
4      tr|AOA075B5P3|AOA075B5P3_MOUSE Immunoglobulin ...      3

```

```

Peptides ... Potential contaminant id Peptide IDs \
0      1 ... NaN 0      9939
1      2 ... NaN 1      5371;7221
2      1 ... NaN 2      7144
3      2 ... NaN 3      3664;23181
4      3 ... NaN 4      2087;3907;22401

```

```

Peptide is razor Mod. peptide IDs \
0      True      10517

```

1	True;True	5685;7650
2	True	7567
3	True;True	3880;3881;24763
4	True;True;True	2218;4134;23956

	Evidence IDs \
0	104111;104112;104113;104114
1	57553;57554;57555;74694;74695
2	74067;74068;74069;74070;74071
3	41615;41616;41617;41618;41619;239742
4	24252;24253;24254;24255;24256;24257;24258;4407...

	MS/MS IDs	Best MS/MS \
0	113199;113200;113201;113202;113203	113200
1	62421;62422;62423;81010;81011	62422;81011
2	80341;80342;80343;80344;80345;80346;80347	80344
3	45159;45160;45161;45162;45163;260731	45163;260731
4	26297;26298;26299;26300;26301;26302;26303;2630...	26297;47805;252619

	Oxidation (M) site IDs	Oxidation (M) site positions
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	0	4
4	NaN	NaN

[5 rows x 371 columns]

[6]: pGroups.tail()

	Protein IDs \
3755	REV__Q9QZE7
3756	S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5
3757	S4R192
3758	V9GWV8;Q3V335;V9GXX3
3759	V9GX06;AOA0R4J0X7;S4R2G5;Q64467

	Majority protein IDs	Peptide counts (all) \
3755	REV__Q9QZE7	1
3756	S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	16;16;15;15;15;15
3757	S4R192	3
3758	V9GWV8;Q3V335;V9GXX3	4;3;2
3759	V9GX06	4;1;1;1

	Peptide counts (razor+unique)	Peptide counts (unique)	Protein names \
3755	1	1	NaN
3756	1;1;1;1;1;1	1;1;1;1;1;1	NaN
3757	1	1	NaN

3758	1;0;0	1;0;0	NaN
3759	1;1;1;1	1;1;1;1	NaN

	Gene names	Fasta headers \
3755	NaN	sp Q9QZE7 TSNAX_MOUSE Translin-associated prot...
3756	Ank3	tr S4R2J6 S4R2J6_MOUSE Ankyrin-3 OS=Mus muscul...
3757	Sec31a	tr S4R192 S4R192_MOUSE Protein transport prote...
3758	Nedd4	tr V9GWV8 V9GWV8_MOUSE E3 ubiquitin-protein li...
3759	Gm11214	tr V9GX06 V9GX06_MOUSE Predicted gene 11214 (F...

	Number of proteins	Peptides	...	Potential contaminant	id \
3755	1	1	...	NaN	3755
3756	6	16	...	NaN	3756
3757	1	3	...	NaN	3757
3758	3	4	...	NaN	3758
3759	4	4	...	NaN	3759

	Peptide IDs \
3755	13511
3756	1455;1705;2665;4244;5703;9994;10028;12460;1246...
3757	10933;16514;22778
3758	3286;20769;22815;23095
3759	1045;15252;22879;22880

	Peptide is razor \
3755	True
3756	False;True;False;False;False;False;False;False...
3757	False;False;True
3758	False;True;False;False
3759	True;False;False;False

	Mod. peptide IDs \
3755	14418
3756	1533;1796;2831;4501;6036;10573;10608;13162;131...
3757	11562;17664;24344
3758	3485;22231;24385;24386;24676
3759	1093;16340;24452;24453;24454

	Evidence IDs \
3755	136224;136225
3756	16960;16961;16962;16963;16964;16965;16966;1696...
3757	113810;113811;113812;113813;113814;113815;1138...
3758	37878;37879;37880;37881;37882;37883;37884;3788...
3759	12247;12248;154168;154169;154170;154171;154172...

	MS/MS IDs \
3755	147817;147818

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3756 18407;18408;18409;18410;18411;18412;18413;1841...
3757 123686;123687;123688;123689;123690;123691;1236...
3758 41143;41144;41145;41146;41147;41148;41149;4115...
3759 13285;13286;13287;167332;167333;167334;167335;...

```

```

Best MS/MS \
3755 147817
3756 18413;21709;33621;51114;66447;113651;113977;13...
3757 123694;178974;256766
3758 41143;230341;257087;259963
3759 13287;167340;257908;257916

```

```

Oxidation (M) site IDs Oxidation (M) site positions
3755 1946;1947 91;92
3756 552;553 182;203
3757 NaN NaN
3758 936 111
3759 172;173 176;179

```

[5 rows x 371 columns]

```

[7]: #remove proteins identified as "reverse", "only identified by site", and
      → "potential contaminant"
pGroups_proc = pGroups[pGroups.Reverse != '+']
pGroups_proc = pGroups_proc[pGroups_proc['Only identified by site'] != '+']
pGroups_proc = pGroups_proc[pGroups_proc['Potential contaminant'] != '+']
pGroups_proc = pGroups_proc.reset_index(drop=True)
pGroups_proc.tail()

```

```

[7]: Protein IDs \
3608 Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2
3609 Q9Z2Z6
3610 Q9Z315;A0A494B9E9
3611 S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5
3612 S4R192

```

```

Majority protein IDs Peptide counts (all) \
3608 Q9Z2X1;J3QMT0;J3QM80 10;7;7;3;2;1
3609 Q9Z2Z6 3
3610 Q9Z315 7;3
3611 S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5 16;16;15;15;15;15
3612 S4R192 3

```

```

Peptide counts (razor+unique) Peptide counts (unique) \
3608 8;7;7;3;2;1 8;7;7;3;2;1
3609 3 3
3610 7;3 7;3
3611 1;1;1;1;1;1 1;1;1;1;1;1

```

3612

1

1

	Protein names	Gene names	\
3608	Heterogeneous nuclear ribonucleoprotein F;Hete...	Hnrnpf	
3609	Mitochondrial carnitine/acylcarnitine carrier ...	Slc25a20	
3610	U4/U6.U5 tri-snRNP-associated protein 1	Sart1	
3611	NaN	Ank3	
3612	NaN	Sec31a	

	Fasta headers	Number of proteins	\
3608	sp Q9Z2X1 HNRPF_MOUSE Heterogeneous nuclear ri...	6	
3609	sp Q9Z2Z6 MCAT_MOUSE Mitochondrial carnitine/a...	1	
3610	sp Q9Z315 SNUT1_MOUSE U4/U6.U5 tri-snRNP-assoc...	2	
3611	tr S4R2J6 S4R2J6_MOUSE Ankyrin-3 OS=Mus muscul...	6	
3612	tr S4R192 S4R192_MOUSE Protein transport prote...	1	

	Peptides	...	Potential contaminant	id	\
3608	10	...	NaN	3705	
3609	3	...	NaN	3706	
3610	7	...	NaN	3707	
3611	16	...	NaN	3756	
3612	3	...	NaN	3757	

	Peptide IDs	\
3608	2655;9429;10400;13472;13473;15976;19341;19692;...	
3609	418;5649;19235	
3610	398;2538;2539;4589;4689;13650;15713	
3611	1455;1705;2665;4244;5703;9994;10028;12460;1246...	
3612	10933;16514;22778	

	Peptide is razor	\
3608	False;True;True;True;True;True;True;True;False...	
3609	True;True;True	
3610	True;True;True;True;True;True;True	
3611	False;True;False;False;False;False;False;False...	
3612	False;False;True	

	Mod. peptide IDs	\
3608	2820;9982;11005;14366;14367;14368;17104;20682;...	
3609	431;5977;20568	
3610	411;2693;2694;4863;4970;4971;14610;16826	
3611	1533;1796;2831;4501;6036;10573;10608;13162;131...	
3612	11562;17664;24344	

	Evidence IDs	\
3608	30975;30976;30977;30978;30979;30980;30981;3098...	
3609	5063;5064;5065;5066;5067;5068;5069;5070;5071;5...	

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3610 4882;4883;4884;4885;4886;4887;29644;29645;2964...
3611 16960;16961;16962;16963;16964;16965;16966;1696...
3612 113810;113811;113812;113813;113814;113815;1138...

```

```

MS/MS IDs \
3608 33513;33514;33515;33516;33517;33518;33519;3352...
3609 5497;5498;5499;5500;5501;5502;5503;5504;5505;5...
3610 5301;5302;5303;5304;5305;5306;32096;32097;3209...
3611 18407;18408;18409;18410;18411;18412;18413;1841...
3612 123686;123687;123688;123689;123690;123691;1236...

```

```

Best MS/MS \
3608 33513;107092;117607;147535;147570;174314;21329...
3609 5506;65881;211665
3610 5304;32096;32124;54448;55730;149357;172113
3611 18413;21709;33621;51114;66447;113651;113977;13...
3612 123694;178974;256766

```

```

Oxidation (M) site IDs Oxidation (M) site positions
3608 1926;1927 2;345
3609 NaN NaN
3610 1928 159
3611 552;553 182;203
3612 NaN NaN

```

[5 rows x 371 columns]

[8]: pGroups\_proc

```

[8]: Protein IDs \
0 G3UZW7;A0A023T778;Q9CQL1;P61327
1 070589;A0A067XG53;F6Y9I5
2 A0A140T8M9;A0A140T8M0;A0A0B4J1I0;A0A075B5N0;A0...
3 A0A075B5M7;A0A0G2JDV4;A0A0B4J1J2;A0A0B4J1J1
4 A0A075B5P3;A0A0A6YVP0;P01867
... ...
3608 Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2
3609 Q9Z2Z6
3610 Q9Z315;A0A494B9E9
3611 S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5
3612 S4R192

```

```

Majority protein IDs Peptide counts (all) \
0 G3UZW7;A0A023T778;Q9CQL1;P61327 1;1;1;1
1 070589;A0A067XG53;F6Y9I5 2;2;1
2 A0A140T8M9;A0A140T8M0;A0A0B4J1I0;A0A075B5N0;A0... 1;1;1;1;1;1;1;1;1
3 A0A075B5M7;A0A0G2JDV4;A0A0B4J1J2;A0A0B4J1J1 2;1;1;1
4 A0A075B5P3;A0A0A6YVP0;P01867 3;3;3

```

...	...	...
3608	Q9Z2X1;J3QMT0;J3QM80	10;7;7;3;2;1
3609	Q9Z2Z6	3
3610	Q9Z315	7;3
3611	S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	16;16;15;15;15;15
3612	S4R192	3

	Peptide counts (razor+unique)	Peptide counts (unique) \
0	1;1;1;1	1;1;1;1
1	2;2;1	2;2;1
2	1;1;1;1;1;1;1;1	1;1;1;1;1;1;1;1
3	2;1;1;1	2;1;1;1
4	3;3;3	3;3;3
...	...	...
3608	8;7;7;3;2;1	8;7;7;3;2;1
3609	3	3
3610	7;3	7;3
3611	1;1;1;1;1;1	1;1;1;1;1;1
3612	1	1

	Protein names \
0	Protein mago nashi homolog 2;Protein mago nash...
1	Peripheral plasma membrane protein CASK
2	Ig kappa chain V-II region 26-10
3	NaN
4	Ig gamma-2B chain C region
...	...
3608	Heterogeneous nuclear ribonucleoprotein F;Hete...
3609	Mitochondrial carnitine/acylcarnitine carrier ...
3610	U4/U6.U5 tri-snRNP-associated protein 1
3611	NaN
3612	NaN

	Gene names \
0	Magohb;Magoh
1	Cask
2	Igkv1-110;Igkv1-35;Igkv1-99;Igkv1-115
3	Igkv5-39;Igkv5-43;Igkv5-45
4	Ighg2b;Igh-3
...	...
3608	Hnrnpf
3609	Slc25a20
3610	Sart1
3611	Ank3
3612	Sec31a

Fasta headers	Number of proteins \
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0	tr G3UZW7 G3UZW7_MOUSE Protein mago nashi homo...	4
1	sp 070589 CSKP_MOUSE Peripheral plasma membran...	3
2	tr AOA140T8M9 AOA140T8M9_MOUSE Immunoglobulin ...	9
3	tr AOA075B5M7 AOA075B5M7_MOUSE Immunoglobulin ...	4
4	tr AOA075B5P3 AOA075B5P3_MOUSE Immunoglobulin ...	3
...	...	...
3608	sp Q9Z2X1 HNRPF_MOUSE Heterogeneous nuclear ri...	6
3609	sp Q9Z2Z6 MCAT_MOUSE Mitochondrial carnitine/a...	1
3610	sp Q9Z315 SNUT1_MOUSE U4/U6.U5 tri-snRNP-assoc...	2
3611	tr S4R2J6 S4R2J6_MOUSE Ankyrin-3 OS=Mus muscul...	6
3612	tr S4R192 S4R192_MOUSE Protein transport prote...	1

	Peptides	...	Potential contaminant	id	\
0	1	...	NaN	0	
1	2	...	NaN	1	
2	1	...	NaN	2	
3	2	...	NaN	3	
4	3	...	NaN	4	
...	...	...	...	...	
3608	10	...	NaN	3705	
3609	3	...	NaN	3706	
3610	7	...	NaN	3707	
3611	16	...	NaN	3756	
3612	3	...	NaN	3757	

	Peptide IDs	\
0	9939	
1	5371;7221	
2	7144	
3	3664;23181	
4	2087;3907;22401	
...	...	
3608	2655;9429;10400;13472;13473;15976;19341;19692;...	
3609	418;5649;19235	
3610	398;2538;2539;4589;4689;13650;15713	
3611	1455;1705;2665;4244;5703;9994;10028;12460;1246...	
3612	10933;16514;22778	

	Peptide is razor	\
0	True	
1	True;True	
2	True	
3	True;True	
4	True;True;True	
...	...	
3608	False;True;True;True;True;True;True;True;False...	
3609	True;True;True	

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3610          True;True;True;True;True;True;True
3611 False;True;False;False;False;False;False...
3612          False;False;True

```

```

Mod. peptide IDs \
0          10517
1          5685;7650
2          7567
3          3880;3881;24763
4          2218;4134;23956
...
3608 2820;9982;11005;14366;14367;14368;17104;20682;...
3609          431;5977;20568
3610          411;2693;2694;4863;4970;4971;14610;16826
3611 1533;1796;2831;4501;6036;10573;10608;13162;131...
3612          11562;17664;24344

```

```

Evidence IDs \
0          104111;104112;104113;104114
1          57553;57554;57555;74694;74695
2          74067;74068;74069;74070;74071
3          41615;41616;41617;41618;41619;239742
4 24252;24253;24254;24255;24256;24257;24258;4407...
...
3608 30975;30976;30977;30978;30979;30980;30981;3098...
3609 5063;5064;5065;5066;5067;5068;5069;5070;5071;5...
3610 4882;4883;4884;4885;4886;4887;29644;29645;2964...
3611 16960;16961;16962;16963;16964;16965;16966;1696...
3612 113810;113811;113812;113813;113814;113815;1138...

```

```

MS/MS IDs \
0          113199;113200;113201;113202;113203
1          62421;62422;62423;81010;81011
2          80341;80342;80343;80344;80345;80346;80347
3          45159;45160;45161;45162;45163;260731
4 26297;26298;26299;26300;26301;26302;26303;2630...
...
3608 33513;33514;33515;33516;33517;33518;33519;3352...
3609 5497;5498;5499;5500;5501;5502;5503;5504;5505;5...
3610 5301;5302;5303;5304;5305;5306;32096;32097;3209...
3611 18407;18408;18409;18410;18411;18412;18413;1841...
3612 123686;123687;123688;123689;123690;123691;1236...

```

```

Best MS/MS \
0          113200
1          62422;81011
2          80344

```

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3                                45163;260731
4                                26297;47805;252619
...
3608  33513;107092;117607;147535;147570;174314;21329...
3609                                5506;65881;211665
3610          5304;32096;32124;54448;55730;149357;172113
3611  18413;21709;33621;51114;66447;113651;113977;13...
3612                                123694;178974;256766

```

	Oxidation (M) site IDs	Oxidation (M) site positions
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	0	4
4	NaN	NaN
...	...	...
3608	1926;1927	2;345
3609	NaN	NaN
3610	1928	159
3611	552;553	182;203
3612	NaN	NaN

[3613 rows x 371 columns]

```

[9]: #create database of all proteins in protein groups
proteins = pGroups_proc[' Protein IDs'].str.split(";", expand=True)
proteins = proteins.stack()
proteins = list(set(proteins))
proteins = pd.DataFrame(proteins)
proteins.to_csv('outs/' + date + '_allProteinsFromPGroups.csv')

```

```

[94]: #extract LFQ intensity and create new df with expt classifiers
      #instead of group by region or timepoint, add region+timepoint as factors

def make_lfq(dataframe, species):
    temp = dataframe.columns
    lfq = pd.DataFrame()
    filename = str('/Users/kavyasharman/Documents/Projects/microLESA/outs/' +
    ↪date + '_' + species + '_lfq.csv')
    string = "LFQ"
    for i in temp:
        if string in i:
            lfq = lfq.append(dataframe[i])
    lfq.columns = dataframe[' Protein IDs']
    lfq = lfq.reset_index()
    lfq['index'] = lfq['index'].str.strip("LFQ intensity ")
    lfq[['TimePoint', 'BioRep', 'Region', 'TechRep']] = lfq['index'].str.
    ↪split("_", expand=True)

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    lfq['Region'] = lfq['Region'].replace({'LE': 'interface', 'AB': 'SAC', 'CO':
→ 'cortex'})
    lfq['DPI_Region_BioRep'] = lfq['TimePoint'] + "_" + lfq['Region'] + "_" +
→lfq['BioRep']
    lfq['BioRep_TechRep'] = lfq['BioRep'] + "_" + lfq['TechRep']
    lfq['DPI_Region'] = lfq['TimePoint'] + "_" + lfq['Region']
    lfq['index'] = lfq['TimePoint'] + "_" + lfq['Region'] + "_" + lfq['BioRep']
→+ "_" + lfq['TechRep']
    lfq = lfq.set_index('index')
    lfq = lfq.sort_values(by=['TimePoint', 'Region', 'BioRep', 'TechRep'],
→ascending=False)
    # lfq.to_csv(filename)
    return lfq

#extract LFQ intensity for all proteins and create new df 'lfq' with expt
→classifiers
lfq = make_lfq(pGroups_proc, 'combined')
lfq.shape

```

[94]: (42, 3620)

[95]: lfq.head()

```

[95]: Protein IDs      G3UZW7;A0A023T778;Q9CQL1;P61327  070589;A0A067XG53;F6Y9I5  \
index
4DPI_interface_2_4                0.0                0.0
4DPI_interface_2_2                0.0                0.0
4DPI_interface_2_1                0.0                0.0
4DPI_interface_1_2                0.0                0.0
4DPI_interface_1_1                0.0                0.0

Protein IDs      A0A140T8M9;A0A140T8M0;A0A0B4J1I0;A0A075B5N0;A0A075B5K8;F6XWB
2;A0A140T8N1;A0A0G2JDE5;P01631  \
index
4DPI_interface_2_4                0.0
4DPI_interface_2_2                0.0
4DPI_interface_2_1                0.0
4DPI_interface_1_2                0.0
4DPI_interface_1_1                0.0

Protein IDs      A0A075B5M7;A0A0G2JDV4;A0A0B4J1J2;A0A0B4J1J1  \
index
4DPI_interface_2_4                0.0
4DPI_interface_2_2                0.0
4DPI_interface_2_1                0.0
4DPI_interface_1_2                0.0
4DPI_interface_1_1                0.0

```

Protein IDs	AOA075B5P3;AOA0A6YVP0;P01867 \		
index			
4DPI_interface_2_4	0.0		
4DPI_interface_2_2	0.0		
4DPI_interface_2_1	0.0		
4DPI_interface_1_2	0.0		
4DPI_interface_1_1	0.0		
Protein IDs	AOA075B5P5;AOA1Y7VJN6;P03987 \		
index			
4DPI_interface_2_4	0.0		
4DPI_interface_2_2	0.0		
4DPI_interface_2_1	0.0		
4DPI_interface_1_2	0.0		
4DPI_interface_1_1	0.0		
Protein IDs	AOA075B6A0;AOA075B5P6;P01872 \		
index			
4DPI_interface_2_4	11850000.0		
4DPI_interface_2_2	13198000.0		
4DPI_interface_2_1	14041000.0		
4DPI_interface_1_2	8327300.0		
4DPI_interface_1_1	9088300.0		
Protein IDs	AOA075B5R2;AOA075B5R3;AOA0A6YX91;P01786 \		
index			
4DPI_interface_2_4	489990.0		
4DPI_interface_2_2	0.0		
4DPI_interface_2_1	1280300.0		
4DPI_interface_1_2	0.0		
4DPI_interface_1_1	0.0		
Protein IDs	AOA075B5S5;AOA0A6YXL5 AOA075B5T3;J3QK03;AOA0A6YWS9 ... \		
index	...		
4DPI_interface_2_4	0.0	0.0	...
4DPI_interface_2_2	0.0	0.0	...
4DPI_interface_2_1	0.0	0.0	...
4DPI_interface_1_2	0.0	0.0	...
4DPI_interface_1_1	0.0	0.0	...
Protein IDs	Q9Z315;AOA494B9E9 \		
index			
4DPI_interface_2_4	954220.0		
4DPI_interface_2_2	1377400.0		
4DPI_interface_2_1	922440.0		
4DPI_interface_1_2	0.0		
4DPI_interface_1_1	1117600.0		

Protein IDs	S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	S4R192	\
index			
4DPI_interface_2_4		0.0	0.0
4DPI_interface_2_2		0.0	0.0
4DPI_interface_2_1		0.0	0.0
4DPI_interface_1_2		0.0	0.0
4DPI_interface_1_1		0.0	0.0

Protein IDs	TimePoint	BioRep	Region	TechRep	DPI_Region_BioRep	\
index						
4DPI_interface_2_4	4DPI	2	interface	4	4DPI_interface_2	
4DPI_interface_2_2	4DPI	2	interface	2	4DPI_interface_2	
4DPI_interface_2_1	4DPI	2	interface	1	4DPI_interface_2	
4DPI_interface_1_2	4DPI	1	interface	2	4DPI_interface_1	
4DPI_interface_1_1	4DPI	1	interface	1	4DPI_interface_1	

Protein IDs	BioRep_TechRep	DPI_Region
index		
4DPI_interface_2_4	2_4	4DPI_interface
4DPI_interface_2_2	2_2	4DPI_interface
4DPI_interface_2_1	2_1	4DPI_interface
4DPI_interface_1_2	1_2	4DPI_interface
4DPI_interface_1_1	1_1	4DPI_interface

[5 rows x 3620 columns]

```
[96]: # calculate missing values (i.e. LFQ = 0)
lfq_nan = lfq.drop(['TimePoint', 'BioRep', 'Region', 'TechRep',
                    'DPI_Region_BioRep', 'BioRep_TechRep', 'DPI_Region'], axis=1)
lfq_nan = lfq_nan.replace(0, np.nan)
number_missing_values = lfq_nan.transpose().isnull().sum()
percent_missing_values = lfq_nan.transpose().isnull().mean() * 100
percent_missing_values
```

```
[96]: index
4DPI_interface_2_4    52.809300
4DPI_interface_2_2    47.329089
4DPI_interface_2_1    52.421810
4DPI_interface_1_2    77.359535
4DPI_interface_1_1    59.811791
4DPI_cortex_3_3       56.545807
4DPI_cortex_3_2       60.891226
4DPI_cortex_2_3       57.846665
4DPI_cortex_2_2       53.390534
4DPI_cortex_2_1       58.815389
4DPI_cortex_1_2       65.292001
4DPI_cortex_1_1       50.429006
```

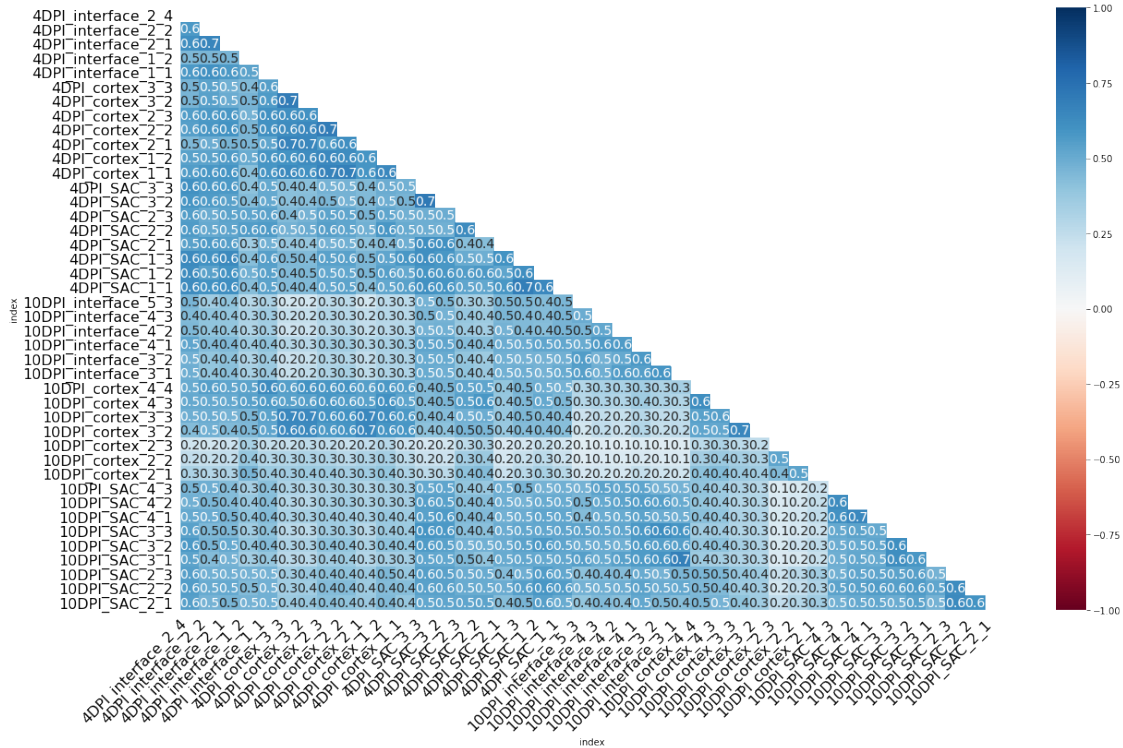
4DPI_SAC_3_3	46.277332
4DPI_SAC_3_2	48.879048
4DPI_SAC_2_3	71.851647
4DPI_SAC_2_2	73.124827
4DPI_SAC_2_1	37.032937
4DPI_SAC_1_3	46.000554
4DPI_SAC_1_2	59.562690
4DPI_SAC_1_1	48.851370
10DPI_interface_5_3	31.082203
10DPI_interface_4_3	38.998063
10DPI_interface_4_2	36.424024
10DPI_interface_4_1	46.360365
10DPI_interface_3_2	36.534736
10DPI_interface_3_1	37.337393
10DPI_cortex_4_4	64.738445
10DPI_cortex_4_3	72.349848
10DPI_cortex_3_3	63.963465
10DPI_cortex_3_2	59.009134
10DPI_cortex_2_3	95.599225
10DPI_cortex_2_2	93.052865
10DPI_cortex_2_1	89.593136
10DPI_SAC_4_3	41.821201
10DPI_SAC_4_2	47.993357
10DPI_SAC_4_1	51.868254
10DPI_SAC_3_3	42.402436
10DPI_SAC_3_2	52.920011
10DPI_SAC_3_1	41.129255
10DPI_SAC_2_3	63.520620
10DPI_SAC_2_2	55.438694
10DPI_SAC_2_1	66.343759

dtype: float64

```
[97]: msno.matrix(lfq_nan.transpose(), sparkline=False, figsize=(20,15))
plt.savefig('outs/' + date + '_missingValues_matrix.png')
```







[17]: *#z-score calculation to drop outliers*

```
z = np.abs(stats.zscore(percent_missing_values))
print(z)
print(np.where(z > 2))
```

```
[0.21484596 0.58111471 0.24074375 1.42596406 0.25316412 0.03488274
 0.32530796 0.12182532 0.17599927 0.1865698 0.61943287 0.37393238
 0.65140871 0.47752355 1.05784547 1.14293821 1.269256 0.66990713
 0.23651554 0.47937339 1.66697207 1.13791721 1.30995253 0.64585919
 1.30255316 1.24890774 0.58243603 1.09114263 0.53064044 0.19951869
 2.64501006 2.47482458 2.24359431 0.9492333 0.5367185 0.27774059
 0.91038662 0.20744659 0.99547936 0.50104297 0.03911095 0.68972687]
(array([30, 31, 32]),)
```

[18]: *#z-score calculation to drop outliers*

```
z = np.abs(stats.zscore(percent_missing_values))
z>2
```

```
[18]: array([False, False, False, False, False, False, False, False, False,
 False, False, False, False, False, False, False, False, False, False,
 False, False, False, True, True, True, False, False, False,
 False, False, False, False, False, False])
```

```
[19]: #list of outlier samples
outliers = percent_missing_values.sort_values(ascending=False)[0:3].index
outliers
```

```
[19]: Index(['10DPI_cortex_2_3', '10DPI_cortex_2_2', '10DPI_cortex_2_1'],
dtype='object', name='index')
```

```
[20]: #drop outliers
combined_lfq = make_lfq(pGroups_proc, 'combined').drop(outliers, axis=0)
combined_lfq.to_csv('outs/' + date + '_combined_lfq.csv')
```

## 0.2 Combined analysis

```
[21]: combined_lfq.head()
```

```
[21]: Protein IDs      G3UZW7;AOA023T778;Q9CQL1;P61327  070589;AOA067XG53;F6Y9I5  \
index
4DPI_interface_2_4                0.0                0.0
4DPI_interface_2_2                0.0                0.0
4DPI_interface_2_1                0.0                0.0
4DPI_interface_1_2                0.0                0.0
4DPI_interface_1_1                0.0                0.0
```

```
Protein IDs      AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA075B5K8;F6XWB
2;AOA140T8N1;AOA0G2JDE5;P01631  \
index
4DPI_interface_2_4                0.0
4DPI_interface_2_2                0.0
4DPI_interface_2_1                0.0
4DPI_interface_1_2                0.0
4DPI_interface_1_1                0.0
```

```
Protein IDs      AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1  \
index
4DPI_interface_2_4                0.0
4DPI_interface_2_2                0.0
4DPI_interface_2_1                0.0
4DPI_interface_1_2                0.0
4DPI_interface_1_1                0.0
```

```
Protein IDs      AOA075B5P3;AOA0A6YVP0;P01867  \
index
4DPI_interface_2_4                0.0
4DPI_interface_2_2                0.0
4DPI_interface_2_1                0.0
4DPI_interface_1_2                0.0
4DPI_interface_1_1                0.0
```

Protein IDs	AOA075B5P5;AOA1Y7VJN6;P03987 \		
index			
4DPI_interface_2_4	0.0		
4DPI_interface_2_2	0.0		
4DPI_interface_2_1	0.0		
4DPI_interface_1_2	0.0		
4DPI_interface_1_1	0.0		

Protein IDs	AOA075B6A0;AOA075B5P6;P01872 \		
index			
4DPI_interface_2_4	11850000.0		
4DPI_interface_2_2	13198000.0		
4DPI_interface_2_1	14041000.0		
4DPI_interface_1_2	8327300.0		
4DPI_interface_1_1	9088300.0		

Protein IDs	AOA075B5R2;AOA075B5R3;AOA0A6YX91;P01786 \		
index			
4DPI_interface_2_4	489990.0		
4DPI_interface_2_2	0.0		
4DPI_interface_2_1	1280300.0		
4DPI_interface_1_2	0.0		
4DPI_interface_1_1	0.0		

Protein IDs	AOA075B5S5;AOA0A6YXL5	AOA075B5T3;J3QK03;AOA0A6YWS9	...	\
index			...	
4DPI_interface_2_4	0.0	0.0	...	
4DPI_interface_2_2	0.0	0.0	...	
4DPI_interface_2_1	0.0	0.0	...	
4DPI_interface_1_2	0.0	0.0	...	
4DPI_interface_1_1	0.0	0.0	...	

Protein IDs	Q9Z315;AOA494B9E9 \		
index			
4DPI_interface_2_4	954220.0		
4DPI_interface_2_2	1377400.0		
4DPI_interface_2_1	922440.0		
4DPI_interface_1_2	0.0		
4DPI_interface_1_1	1117600.0		

Protein IDs	S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	S4R192	\
index			
4DPI_interface_2_4	0.0	0.0	
4DPI_interface_2_2	0.0	0.0	
4DPI_interface_2_1	0.0	0.0	
4DPI_interface_1_2	0.0	0.0	
4DPI_interface_1_1	0.0	0.0	

Protein IDs	TimePoint	BioRep	Region	TechRep	DPI_Region_BioRep \
index					
4DPI_interface_2_4	4DPI	2	interface	4	4DPI_interface_2
4DPI_interface_2_2	4DPI	2	interface	2	4DPI_interface_2
4DPI_interface_2_1	4DPI	2	interface	1	4DPI_interface_2
4DPI_interface_1_2	4DPI	1	interface	2	4DPI_interface_1
4DPI_interface_1_1	4DPI	1	interface	1	4DPI_interface_1

Protein IDs	BioRep_TechRep	DPI_Region
index		
4DPI_interface_2_4	2_4	4DPI_interface
4DPI_interface_2_2	2_2	4DPI_interface
4DPI_interface_2_1	2_1	4DPI_interface
4DPI_interface_1_2	1_2	4DPI_interface
4DPI_interface_1_1	1_1	4DPI_interface

[5 rows x 3620 columns]

```
[22]: #plot venn diagram of protein groups
df = combined_lfq.groupby('Region').sum()

df = df.replace(0, np.nan)
df_missing_values = df.transpose().isnull().sum()
df_present_values = df.shape[1]-df_missing_values

df1 = df.transpose()

df_ab = df1['SAC']
df_ab = df_ab.dropna()
set_ab = df_ab.index.values
set_ab = set(set_ab)

df_le = df1['interface']
df_le = df_le.dropna()
set_le = df_le.index.values
set_le = set(set_le)

df_co = df1['cortex']
df_co = df_co.dropna()
set_co = df_co.index.values
set_co = set(set_co)

df1.head()

df1.dropna(subset=['SAC'])
```

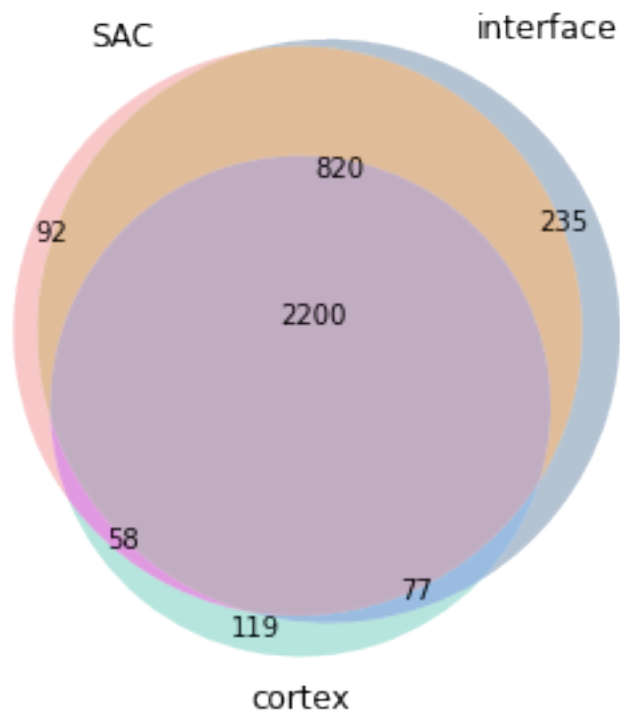
```

#from importlib import reload
#reload(plt)

plt.figure(figsize=(5,5))
vd = venn3([set_ab, set_le, set_co], ('SAC', 'interface', 'cortex'))
vd.get_patch_by_id("100").set_color("#EF767A")
vd.get_patch_by_id('100').set_edgecolor('none')
vd.get_patch_by_id('100').set_alpha(0.4)
vd.get_patch_by_id("001").set_color("#49BEAA")
vd.get_patch_by_id('001').set_edgecolor('none')
vd.get_patch_by_id('001').set_alpha(0.4)
vd.get_patch_by_id("010").set_color("#456990")
vd.get_patch_by_id('010').set_edgecolor('none')
vd.get_patch_by_id('010').set_alpha(0.4)

plt.savefig('outs/' + date + '_venn_combined.svg')

```



```
[29]: ### Function to separate data and labels
def df_sep(label, dataframe):
    X = dataframe.drop(['TimePoint', 'TechRep', 'BioRep', 'Region',
        ↳ 'DPI_Region_BioRep', 'DPI_Region', 'BioRep_TechRep'], axis=1)
    y = dataframe[label]
    return X,y

[30]: X,Y = df_sep('DPI_Region', combined_lfq)
Y
y = pd.factorize(Y)[0]

[31]: X.to_csv('/Users/kavyasharman/Documents/Projects/microLESA/outs/' + date +
    ↳ '_combined_X.csv')
```

### 0.3 PCA

```
[32]: pca = PCA()
pca.fit_transform(X)

[32]: array([[ -1.20613918e+09, -7.35069505e+08,  1.20642660e+09, ...,
        -9.66779606e+06,  3.77862408e+07,  2.94952004e-07],
        [ 5.50887608e+08, -6.44020375e+08,  3.78395103e+07, ...,
        4.48483120e+07,  3.02168233e+07,  2.94952004e-07],
        [ 2.78911665e+08, -1.35151935e+09, -1.10286837e+08, ...,
        -3.92765751e+07, -2.82151875e+07,  2.94952004e-07],
        ...,
        [-2.78629898e+09, -1.91817266e+08,  2.13531506e+08, ...,
        -7.88043640e+06, -1.54576178e+07,  2.94952004e-07],
        [-2.14104411e+09, -9.88192949e+07,  8.11645909e+08, ...,
        -2.20069992e+06, -1.01797990e+07,  2.94952004e-07],
        [-2.01959630e+09, -1.92026814e+08, -3.13821517e+08, ...,
        2.28581628e+06,  2.49944228e+07,  2.94952004e-07]])

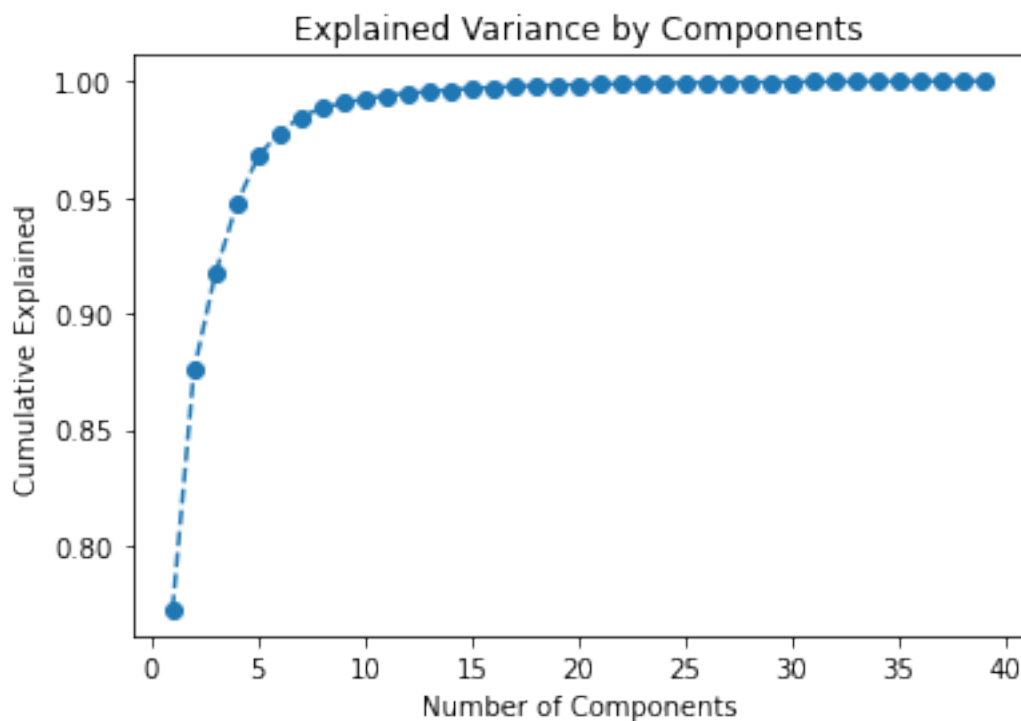
[33]: components = pca.components_
components.shape

[33]: (39, 3613)

[34]: X.shape

[34]: (39, 3613)

[35]: plt.figure()
plt.plot(range(1,40), pca.explained_variance_ratio_.cumsum(), marker = 'o',
    ↳ linestyle = '--')
plt.title('Explained Variance by Components')
plt.xlabel('Number of Components')
plt.ylabel('Cumulative Explained')
plt.savefig('outs/' + date + '_combined_PCA_explainedVariance.svg')
```



```
[36]: pca_expl_var = pca.explained_variance_ratio_
      pca_expl_var
```

```
[36]: array([7.72585327e-01, 1.03009194e-01, 4.14244991e-02, 2.99833868e-02,
          2.06562695e-02, 1.00463271e-02, 6.99901733e-03, 3.76353515e-03,
          2.13194779e-03, 1.71263813e-03, 1.48846065e-03, 9.14578266e-04,
          8.24462531e-04, 6.94519592e-04, 5.66674631e-04, 4.76616717e-04,
          4.02062568e-04, 2.88359638e-04, 2.56272367e-04, 2.26789690e-04,
          2.22757395e-04, 1.81912857e-04, 1.54660794e-04, 1.38718971e-04,
          1.23622374e-04, 1.12507129e-04, 9.61538441e-05, 8.23296203e-05,
          6.69051649e-05, 6.17230689e-05, 5.51396189e-05, 5.09239881e-05,
          4.03391810e-05, 3.97261547e-05, 3.53797674e-05, 3.17609348e-05,
          2.94709459e-05, 2.50296281e-05, 4.93141904e-33])
```

```
[37]: scores_pca = pca.fit_transform(X)

      #pca loadings
      loadings = pd.DataFrame(pca.components_.T, index=X.columns)
```

```
[38]: scores_pca.shape
```

```
[38]: (39, 39)
```

```
[39]: pd.DataFrame(scores_pca).to_csv('/Users/kavyasharman/Documents/Projects/
      ↳microLESA/outs/' + date + '_combined_scores_pca.csv')
```

```
[40]: loadings.to_csv('/Users/kavyasharman/Documents/Projects/microLESA/outs/' + date_
→+ '_combined_loadings_pca.csv')
```

### 0.3.1 Silhouette scores to determine k

```
[41]: from sklearn.datasets import make_blobs
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette_samples, silhouette_score

import matplotlib.pyplot as plt
import matplotlib.cm as cm
import numpy as np

range_n_clusters = [2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15]
silhouette_scores_euc = pd.DataFrame(columns = ['k', 'score'])

for n_clusters in range_n_clusters:
    # Create a subplot with 1 row and 2 columns
    fig, (ax1, ax2) = plt.subplots(1, 2)
    fig.set_size_inches(18, 7)

    # The 1st subplot is the silhouette plot
    # The silhouette coefficient can range from -1, 1 but in this example all
    # lie within [-0.1, 1]
    ax1.set_xlim([-0.1, 1])
    # The (n_clusters+1)*10 is for inserting blank space between silhouette
    # plots of individual clusters, to demarcate them clearly.
    ax1.set_ylim([0, len(scores_pca) + (n_clusters + 1) * 10])

    # Initialize the clusterer with n_clusters value and a random generator
    # seed of 10 for reproducibility.
    clusterer = KMeans(n_clusters=n_clusters, random_state=42)
    cluster_labels = clusterer.fit_predict(scores_pca)

    # The silhouette_score gives the average value for all the samples.
    # This gives a perspective into the density and separation of the formed
    # clusters
    silhouette_avg = silhouette_score(scores_pca, cluster_labels)
    # print("For n_clusters =", n_clusters,
    #       "The average silhouette_score is :", silhouette_avg)
    silhouette_scores_euc = silhouette_scores_euc.append(pd.DataFrame({'k':
→n_clusters, 'score': silhouette_avg}, index=[0]), ignore_index=True)

    # Compute the silhouette scores for each sample
    sample_silhouette_values = silhouette_samples(scores_pca, cluster_labels)
```



```

y_lower = 10
for i in range(n_clusters):
    # Aggregate the silhouette scores for samples belonging to
    # cluster i, and sort them
    ith_cluster_silhouette_values = \
        sample_silhouette_values[cluster_labels == i]

    ith_cluster_silhouette_values.sort()

    size_cluster_i = ith_cluster_silhouette_values.shape[0]
    y_upper = y_lower + size_cluster_i

    color = cm.nipy_spectral(float(i) / n_clusters)
    ax1.fill_betweenx(np.arange(y_lower, y_upper),
                      0, ith_cluster_silhouette_values,
                      facecolor=color, edgecolor=color, alpha=0.7)

    # Label the silhouette plots with their cluster numbers at the middle
    ax1.text(-0.05, y_lower + 0.5 * size_cluster_i, str(i))

    # Compute the new y_lower for next plot
    y_lower = y_upper + 10  # 10 for the 0 samples

ax1.set_title("The silhouette plot for the various clusters.")
ax1.set_xlabel("The silhouette coefficient values")
ax1.set_ylabel("Cluster label")

# The vertical line for average silhouette score of all the values
ax1.axvline(x=silhouette_avg, color="red", linestyle="--")

ax1.set_yticks([])  # Clear the yaxis labels / ticks
ax1.set_xticks([-0.1, 0, 0.2, 0.4, 0.6, 0.8, 1])

# 2nd Plot showing the actual clusters formed
colors = cm.nipy_spectral(cluster_labels.astype(float) / n_clusters)
ax2.scatter(scores_pca[:, 0], scores_pca[:, 1], marker='.', s=300, lw=0,
→alpha=0.7,
           c=colors, edgecolor='k')

# Labeling the clusters
centers = clusterer.cluster_centers_
# Draw white circles at cluster centers
ax2.scatter(centers[:, 0], centers[:, 1], marker='o',
           c="white", alpha=1, s=200, edgecolor='k')

for i, c in enumerate(centers):
    ax2.scatter(c[0], c[1], marker='$%d$' % i, alpha=1,

```

```

s=50, edgecolor='k')

ax2.set_title("The visualization of the clustered data.")
ax2.set_xlabel("Feature space for the 1st feature")
ax2.set_ylabel("Feature space for the 2nd feature")

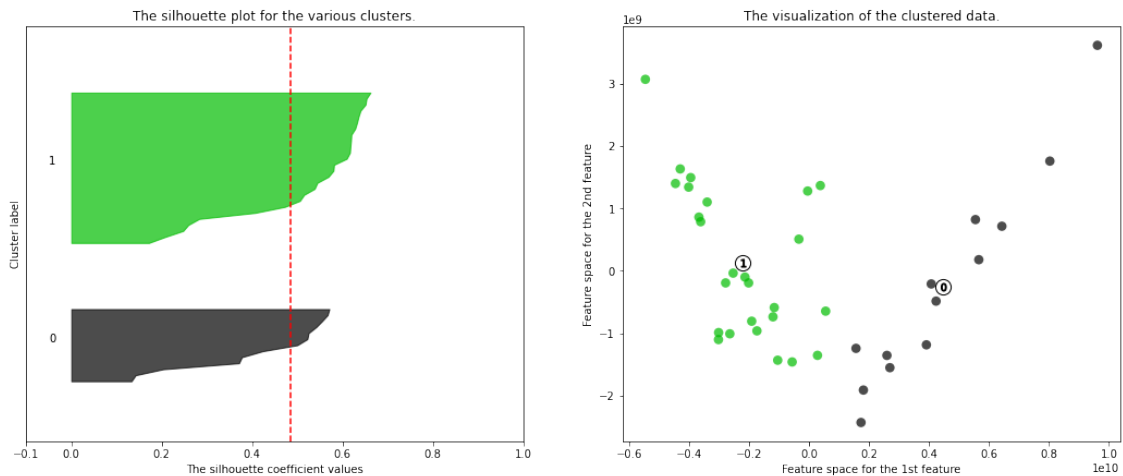
plt.suptitle(("Silhouette analysis for KMeans clustering on sample data "
              "with n_clusters = %d" % n_clusters),
             fontsize=14, fontweight='bold')

plt.savefig('outs/' + date + '_combined_silhouette_scores_euc_' +
            str(n_clusters) + '.svg')

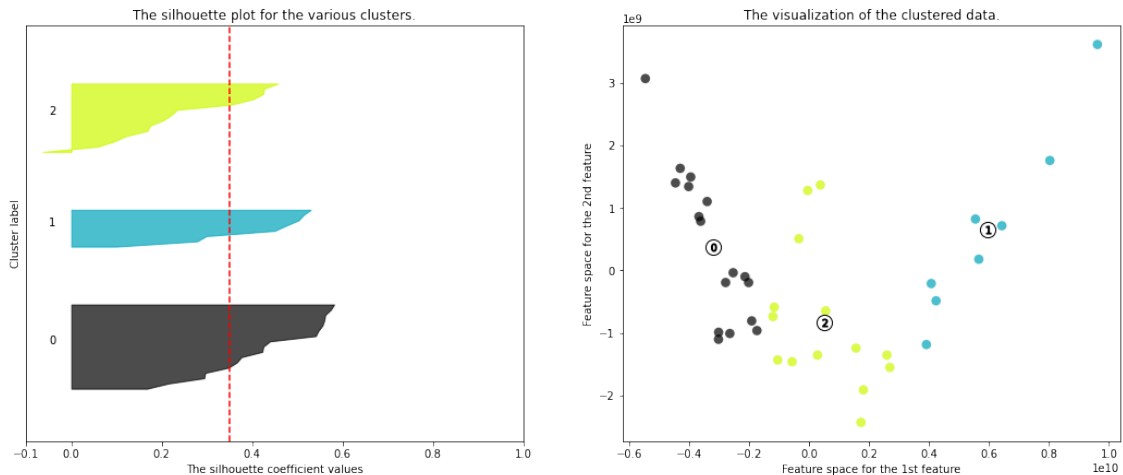
silhouette_scores_euc.to_csv('outs/' + date + '_combined_silhouette_scores_euc.'
                             + csv')

```

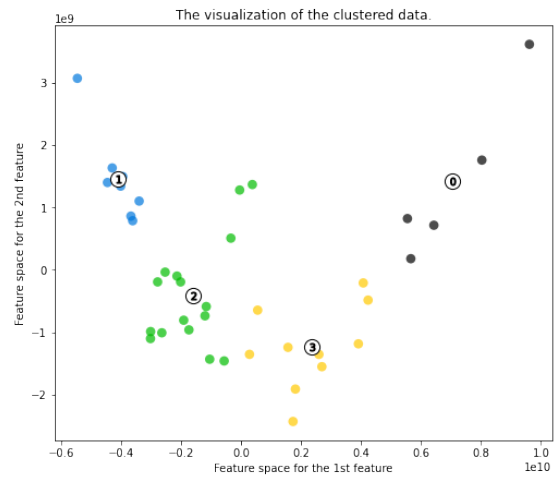
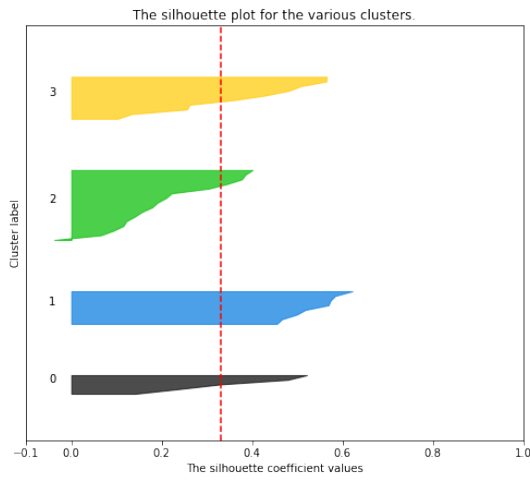
**Silhouette analysis for KMeans clustering on sample data with n\_clusters = 2**



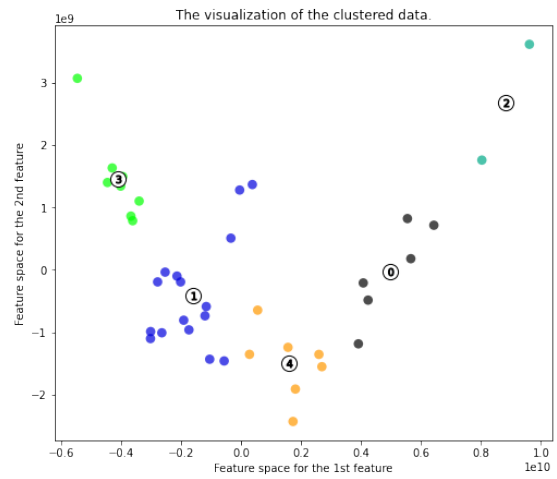
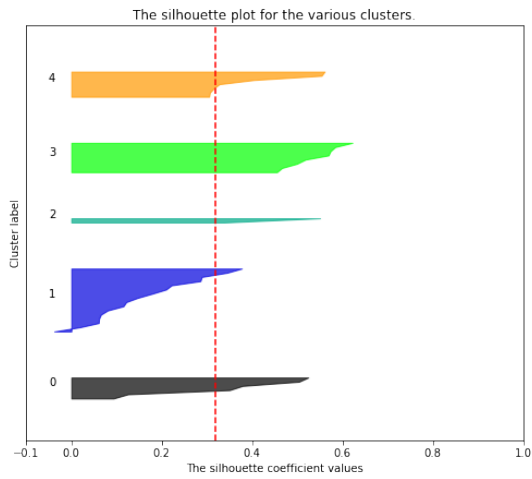
**Silhouette analysis for KMeans clustering on sample data with n\_clusters = 3**



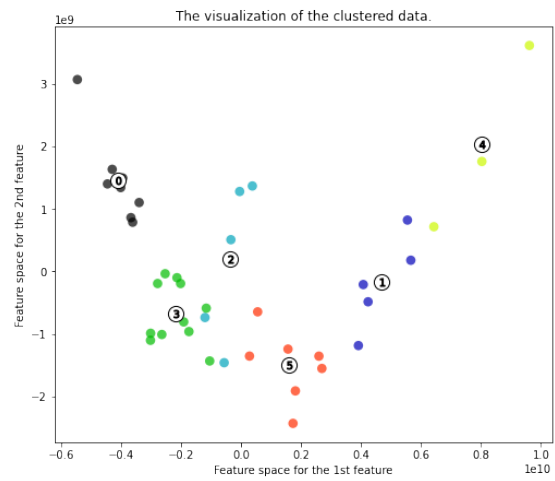
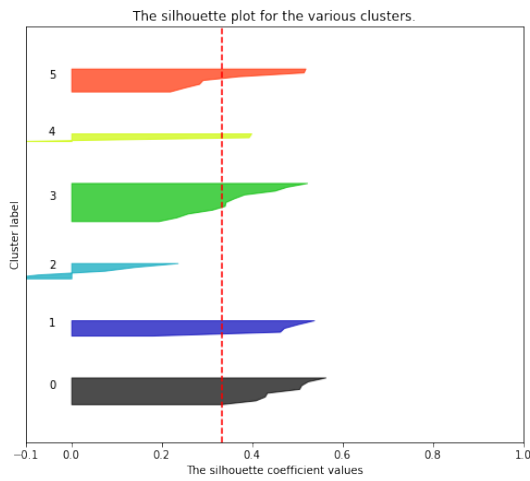
### Silhouette analysis for KMeans clustering on sample data with $n\_clusters = 4$



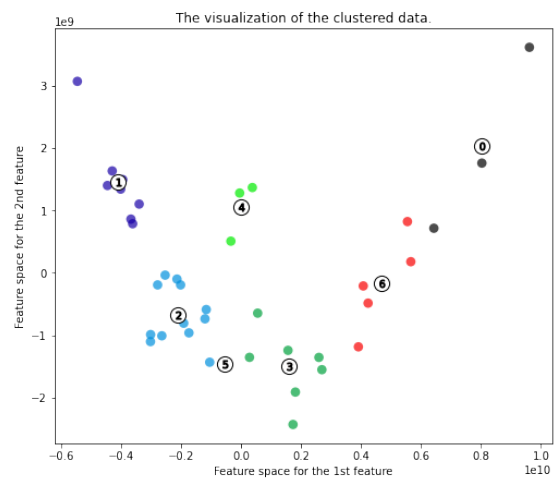
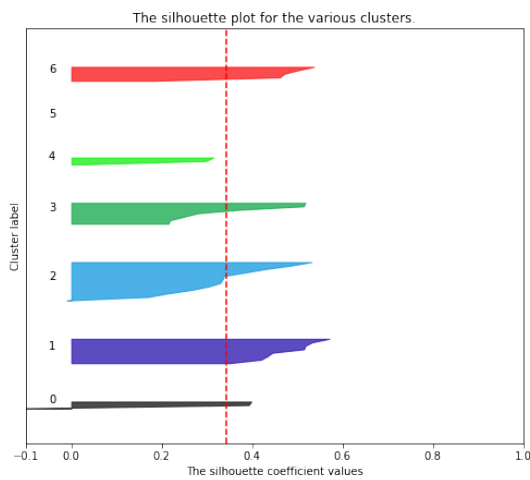
### Silhouette analysis for KMeans clustering on sample data with $n\_clusters = 5$



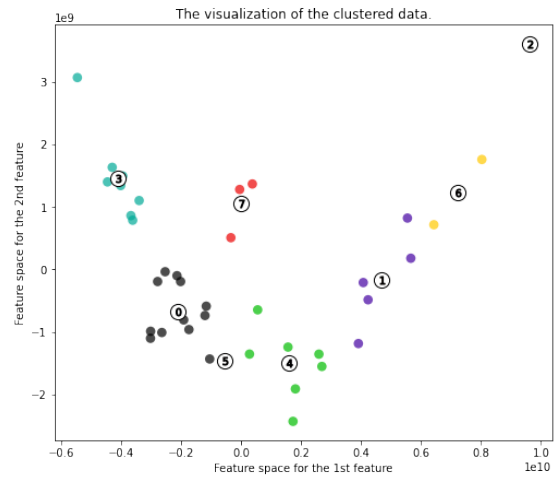
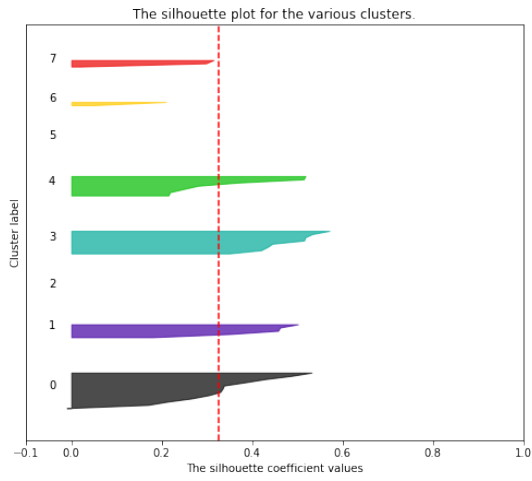
### Silhouette analysis for KMeans clustering on sample data with $n\_clusters = 6$



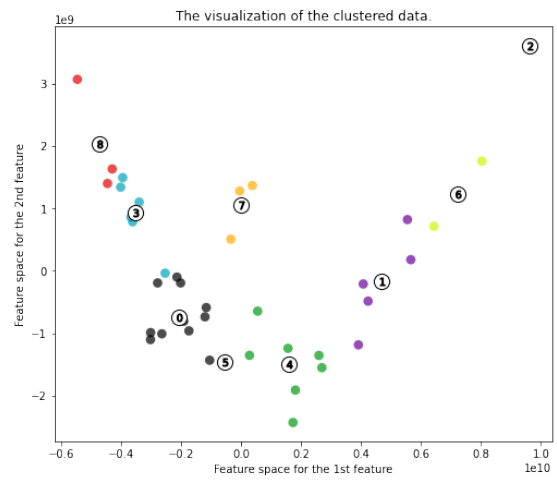
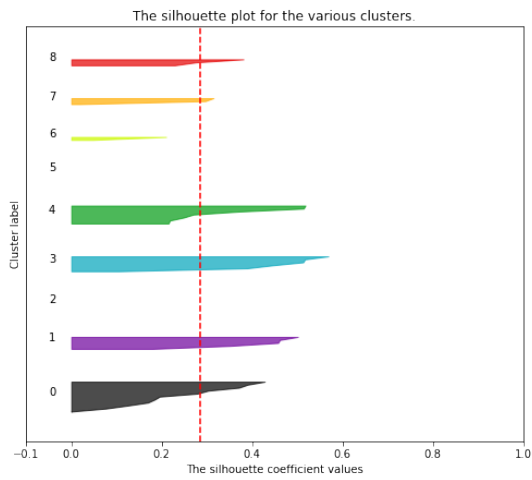
### Silhouette analysis for KMeans clustering on sample data with $n\_clusters = 7$



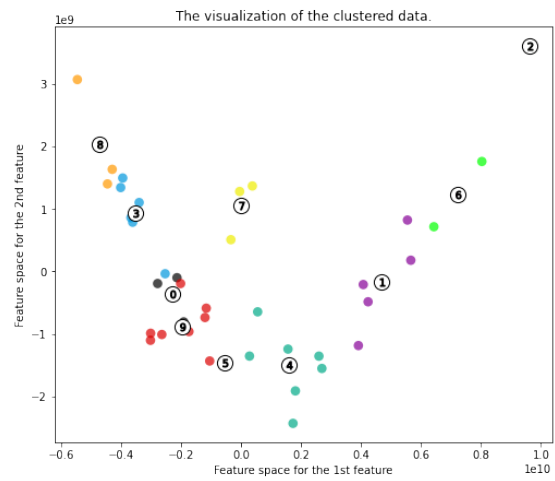
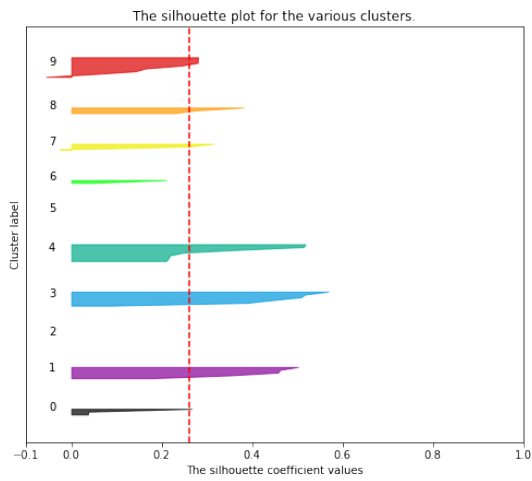
### Silhouette analysis for KMeans clustering on sample data with $n\_clusters = 8$



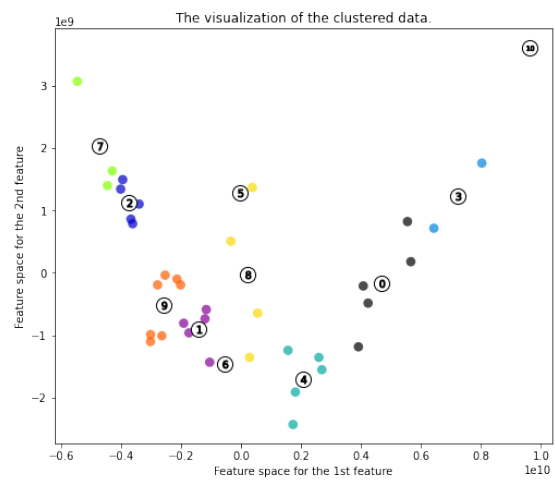
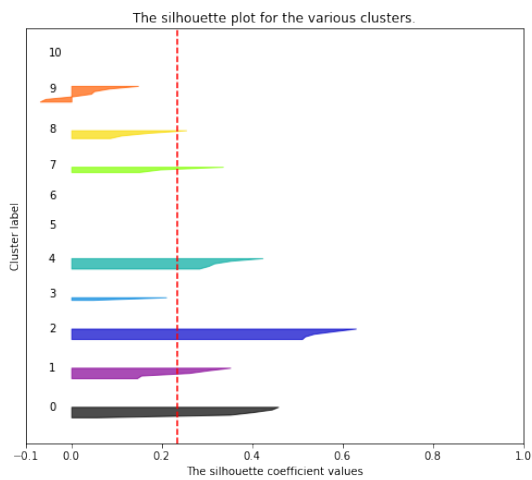
### Silhouette analysis for KMeans clustering on sample data with $n\_clusters = 9$



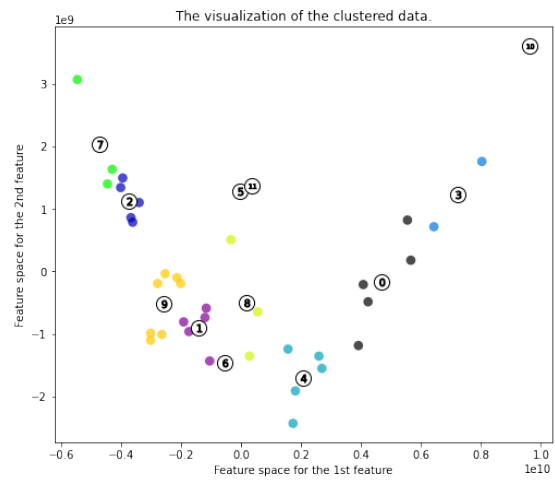
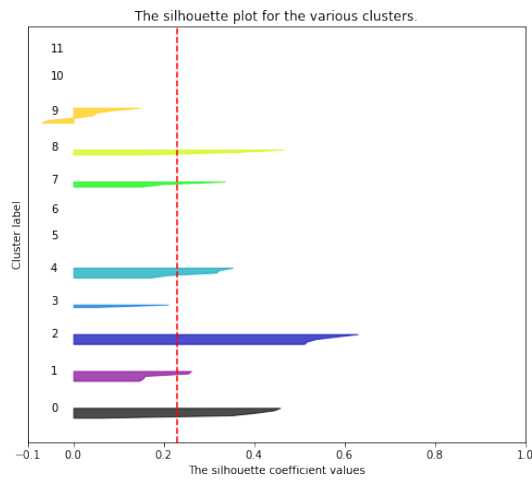
### Silhouette analysis for KMeans clustering on sample data with n\_clusters = 10



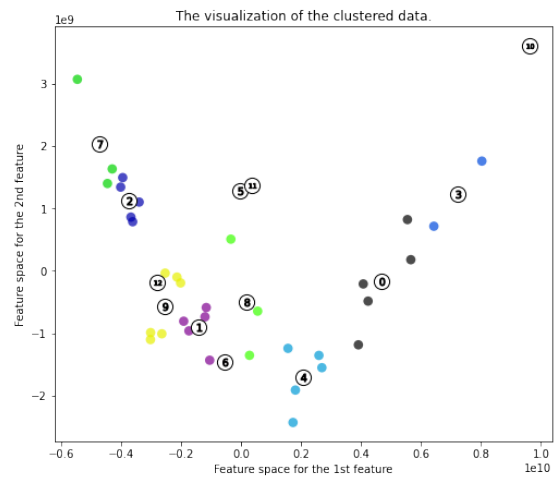
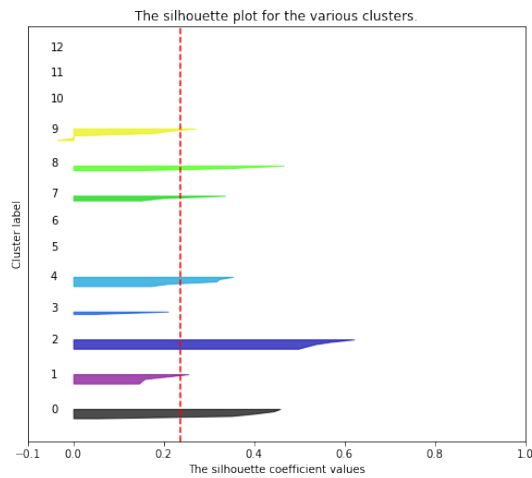
### Silhouette analysis for KMeans clustering on sample data with n\_clusters = 11



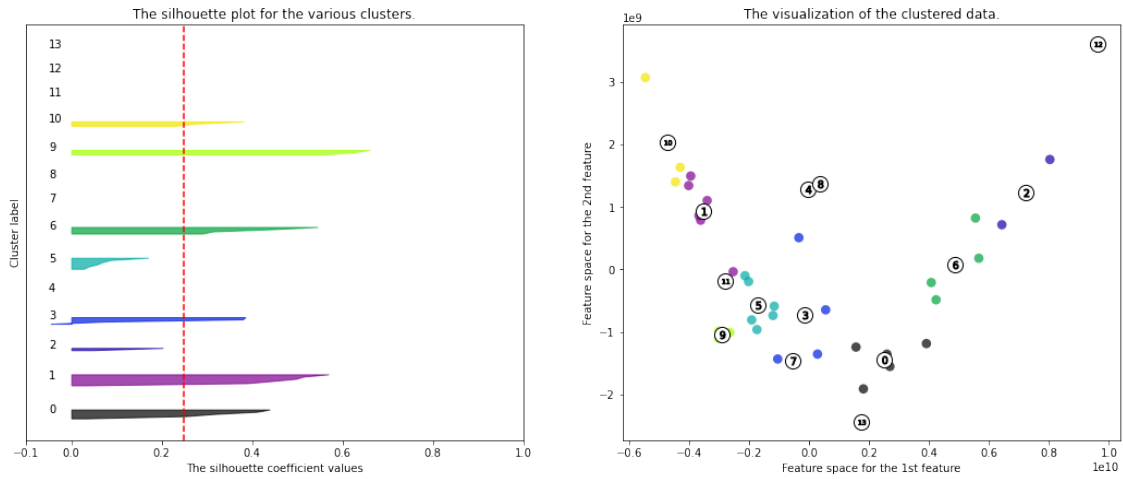
### Silhouette analysis for KMeans clustering on sample data with n\_clusters = 12



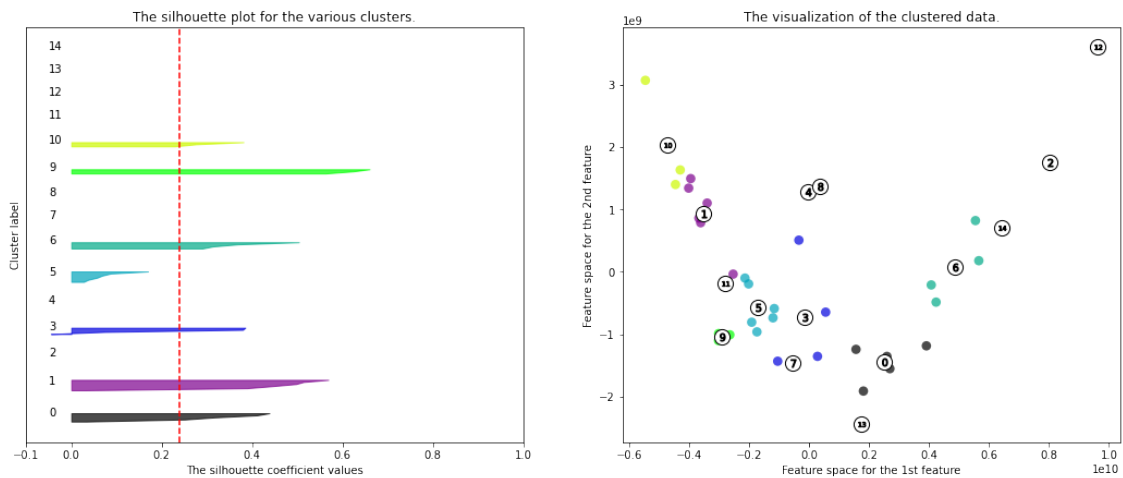
### Silhouette analysis for KMeans clustering on sample data with n\_clusters = 13



### Silhouette analysis for KMeans clustering on sample data with n\_clusters = 14



### Silhouette analysis for KMeans clustering on sample data with n\_clusters = 15



[42]: silhouette\_scores\_euc

[42]:

	k	score
0	2	0.484378
1	3	0.351101
2	4	0.331259
3	5	0.319485
4	6	0.334392
5	7	0.342704
6	8	0.327104
7	9	0.285408
8	10	0.260599
9	11	0.234079



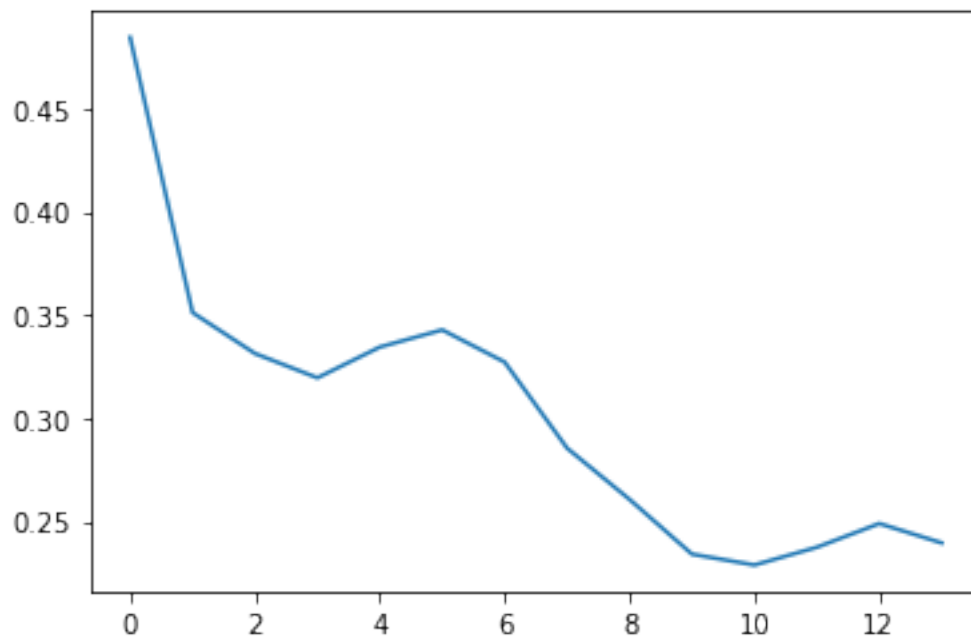
```

10 12 0.228832
11 13 0.237426
12 14 0.248910
13 15 0.239528

```

```
[43]: plt.plot(silhouette_scores_euc['score'])
```

```
[43]: [<matplotlib.lines.Line2D at 0x7facaf0bbfa0>]
```



### 0.3.2 k=5

```
[44]: kmeans_pca = KMeans(n_clusters=5, init='k-means++', random_state=42)
kmeans_pca.fit(scores_pca)
kmeans_labels = kmeans_pca.labels_
```

```
[45]: kmeans_labels_df = pd.DataFrame(kmeans_labels)
```

```
[46]: cluster_centers = kmeans_pca.cluster_centers_
cluster_centers_df = pd.DataFrame(cluster_centers).to_csv('outs/' + date + '_'
↳ '_combined_kmeans_clusterCenters.csv')
```

```
[48]: cluster_centers = kmeans_pca.cluster_centers_
df_segm_pca_kmeans = pd.concat([X.reset_index(drop=True), pd.
↳ DataFrame(scores_pca)], axis=1)
df_segm_pca_kmeans.columns.values[-34]
df_segm_pca_kmeans.columns.values[-39:-34] = ['Component 1', 'Component 2',
↳ 'Component 3', 'Component 4', 'Component 5']
df_segm_pca_kmeans['Segment K-means PCA'] = kmeans_pca.labels_
```

```

df_segmpca_kmeans.index = X.index
df_segmpca_kmeans[['TimePoint', 'BioRep', 'TechRep', 'Region', 'DPI_Region',
    → 'BioRep_TechRep']] = combined_lfq[['TimePoint', 'BioRep', 'TechRep',
    → 'Region', 'DPI_Region', 'BioRep_TechRep']]
df_segmpca_kmeans.to_csv('outs/' + date + '_combined_df_segmpca_kmeans.csv')
df_segmpca_kmeans

```

[48]: G3UZW7;A0A023T778;Q9CQL1;P61327 \

index	
4DPI_interface_2_4	0.0
4DPI_interface_2_2	0.0
4DPI_interface_2_1	0.0
4DPI_interface_1_2	0.0
4DPI_interface_1_1	0.0
4DPI_cortex_3_3	0.0
4DPI_cortex_3_2	0.0
4DPI_cortex_2_3	0.0
4DPI_cortex_2_2	0.0
4DPI_cortex_2_1	0.0
4DPI_cortex_1_2	0.0
4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	189280.0
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4DPI_SAC_1_2	779590.0
4DPI_SAC_1_1	0.0
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10DPI_interface_4_3	0.0
10DPI_interface_4_2	222210.0
10DPI_interface_4_1	0.0
10DPI_interface_3_2	0.0
10DPI_interface_3_1	0.0
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10DPI_cortex_4_3	0.0
10DPI_cortex_3_3	0.0
10DPI_cortex_3_2	0.0
10DPI_SAC_4_3	0.0
10DPI_SAC_4_2	0.0
10DPI_SAC_4_1	0.0
10DPI_SAC_3_3	0.0
10DPI_SAC_3_2	0.0
10DPI_SAC_3_1	0.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	0.0

10DPI\_SAC\_2\_1 0.0

070589;AOA067XG53;F6Y9I5 \

index

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4DPI_cortex_2_2	0.0
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4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
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4DPI_SAC_1_1	0.0
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10DPI_interface_4_3	0.0
10DPI_interface_4_2	0.0
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10DPI_SAC_4_1	0.0
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10DPI_SAC_3_2	0.0
10DPI_SAC_3_1	0.0
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10DPI_SAC_2_2	0.0
10DPI_SAC_2_1	0.0

AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA075B5K8;F6XW

B2;AOA140T8N1;AOA0G2JDE5;P01631 \

index

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4DPI_interface_2_1	0.0
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4DPI_cortex_2_3	0.0
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4DPI_cortex_2_1	0.0
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4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
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10DPI_SAC_3_1	0.0
10DPI_SAC_2_3	0.0
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10DPI_SAC_2_1	0.0

AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1 \

index

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4DPI_interface_1_1	0.0

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4DPI_cortex_2_2	0.0
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4DPI_cortex_1_2	0.0
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10DPI_SAC_2_1	0.0

AOA075B5P3;AOA0A6YVP0;P01867 \

index

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4DPI_cortex_2_1	0.0

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AOA075B5P5;AOA1Y7VJN6;P03987 \

index

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4DPI_cortex_2_1	0.0
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4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0

4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
4DPI_SAC_1_2	0.0
4DPI_SAC_1_1	0.0
10DPI_interface_5_3	1109600.0
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10DPI_interface_4_2	14386000.0
10DPI_interface_4_1	19824000.0
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10DPI_cortex_4_4	21601000.0
10DPI_cortex_4_3	526710.0
10DPI_cortex_3_3	1102700.0
10DPI_cortex_3_2	654170.0
10DPI_SAC_4_3	29513000.0
10DPI_SAC_4_2	13752000.0
10DPI_SAC_4_1	17690000.0
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10DPI_SAC_3_2	997060.0
10DPI_SAC_3_1	295760.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	1941400.0
10DPI_SAC_2_1	0.0

AOA075B6A0;AOA075B5P6;P01872 \

index

4DPI_interface_2_4	11850000.0
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4DPI_cortex_3_2	9475600.0
4DPI_cortex_2_3	12548000.0
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4DPI_cortex_2_1	18376000.0
4DPI_cortex_1_2	13749000.0
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4DPI_SAC_3_3	4038600.0
4DPI_SAC_3_2	1051500.0
4DPI_SAC_2_3	10500000.0
4DPI_SAC_2_2	4068800.0
4DPI_SAC_2_1	6212200.0
4DPI_SAC_1_3	7820100.0
4DPI_SAC_1_2	3859700.0
4DPI_SAC_1_1	4501600.0

10DPI_interface_5_3	7111900.0
10DPI_interface_4_3	15394000.0
10DPI_interface_4_2	20355000.0
10DPI_interface_4_1	24453000.0
10DPI_interface_3_2	6101300.0
10DPI_interface_3_1	9250300.0
10DPI_cortex_4_4	49938000.0
10DPI_cortex_4_3	16417000.0
10DPI_cortex_3_3	16583000.0
10DPI_cortex_3_2	14249000.0
10DPI_SAC_4_3	36836000.0
10DPI_SAC_4_2	21015000.0
10DPI_SAC_4_1	20078000.0
10DPI_SAC_3_3	5357600.0
10DPI_SAC_3_2	2988700.0
10DPI_SAC_3_1	6025900.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	6863300.0
10DPI_SAC_2_1	2568900.0

AOA075B5R2;AOA075B5R3;AOA0A6YX91;P01786 \

index

4DPI_interface_2_4	489990.0
4DPI_interface_2_2	0.0
4DPI_interface_2_1	1280300.0
4DPI_interface_1_2	0.0
4DPI_interface_1_1	0.0
4DPI_cortex_3_3	0.0
4DPI_cortex_3_2	0.0
4DPI_cortex_2_3	537080.0
4DPI_cortex_2_2	1492000.0
4DPI_cortex_2_1	247720.0
4DPI_cortex_1_2	0.0
4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
4DPI_SAC_1_2	0.0
4DPI_SAC_1_1	0.0
10DPI_interface_5_3	923480.0
10DPI_interface_4_3	2742000.0
10DPI_interface_4_2	2199900.0
10DPI_interface_4_1	2193900.0
10DPI_interface_3_2	875140.0



10DPI_interface_3_1	1436900.0
10DPI_cortex_4_4	1717800.0
10DPI_cortex_4_3	1395700.0
10DPI_cortex_3_3	880010.0
10DPI_cortex_3_2	708140.0
10DPI_SAC_4_3	1401500.0
10DPI_SAC_4_2	0.0
10DPI_SAC_4_1	1453700.0
10DPI_SAC_3_3	536050.0
10DPI_SAC_3_2	2143600.0
10DPI_SAC_3_1	944670.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	533920.0
10DPI_SAC_2_1	0.0

	AOA075B5S5;AOA0A6YXL5	AOA075B5T3;J3QK03;AOA0A6YWS9	...	\
index			...	
4DPI_interface_2_4	0.0	0.0	...	
4DPI_interface_2_2	0.0	0.0	...	
4DPI_interface_2_1	0.0	0.0	...	
4DPI_interface_1_2	0.0	0.0	...	
4DPI_interface_1_1	0.0	0.0	...	
4DPI_cortex_3_3	0.0	0.0	...	
4DPI_cortex_3_2	0.0	0.0	...	
4DPI_cortex_2_3	0.0	0.0	...	
4DPI_cortex_2_2	0.0	0.0	...	
4DPI_cortex_2_1	0.0	0.0	...	
4DPI_cortex_1_2	0.0	0.0	...	
4DPI_cortex_1_1	0.0	0.0	...	
4DPI_SAC_3_3	0.0	0.0	...	
4DPI_SAC_3_2	0.0	0.0	...	
4DPI_SAC_2_3	0.0	0.0	...	
4DPI_SAC_2_2	0.0	0.0	...	
4DPI_SAC_2_1	0.0	0.0	...	
4DPI_SAC_1_3	0.0	0.0	...	
4DPI_SAC_1_2	0.0	0.0	...	
4DPI_SAC_1_1	0.0	0.0	...	
10DPI_interface_5_3	582000.0	624130.0	...	
10DPI_interface_4_3	10474000.0	1520300.0	...	
10DPI_interface_4_2	4267600.0	3319600.0	...	
10DPI_interface_4_1	1899800.0	1112100.0	...	
10DPI_interface_3_2	521140.0	826650.0	...	
10DPI_interface_3_1	554440.0	0.0	...	
10DPI_cortex_4_4	3546300.0	2818100.0	...	
10DPI_cortex_4_3	0.0	2965700.0	...	
10DPI_cortex_3_3	0.0	992900.0	...	
10DPI_cortex_3_2	739180.0	0.0	...	

10DPI_SAC_4_3	2447600.0	5502600.0	...
10DPI_SAC_4_2	3972000.0	2063300.0	...
10DPI_SAC_4_1	8814000.0	2556300.0	...
10DPI_SAC_3_3	0.0	1867700.0	...
10DPI_SAC_3_2	0.0	766080.0	...
10DPI_SAC_3_1	0.0	0.0	...
10DPI_SAC_2_3	0.0	0.0	...
10DPI_SAC_2_2	0.0	0.0	...
10DPI_SAC_2_1	0.0	0.0	...

	36	37	38 \
index			
4DPI_interface_2_4	-9.667796e+06	3.778624e+07	2.949520e-07
4DPI_interface_2_2	4.484831e+07	3.021682e+07	2.949520e-07
4DPI_interface_2_1	-3.927658e+07	-2.821519e+07	2.949520e-07
4DPI_interface_1_2	1.602163e+07	1.157634e+07	2.949520e-07
4DPI_interface_1_1	-2.771801e+07	-1.684883e+07	2.949520e-07
4DPI_cortex_3_3	-2.290089e+07	1.351457e+07	2.949520e-07
4DPI_cortex_3_2	1.978959e+07	-2.313470e+07	2.949520e-07
4DPI_cortex_2_3	6.588185e+06	-3.016993e+07	2.949520e-07
4DPI_cortex_2_2	-2.862014e+07	5.637718e+07	2.949520e-07
4DPI_cortex_2_1	3.341529e+05	2.571811e+05	2.949520e-07
4DPI_cortex_1_2	-4.056160e+07	-1.501557e+07	2.949520e-07
4DPI_cortex_1_1	7.268118e+07	-1.174222e+07	2.949520e-07
4DPI_SAC_3_3	-5.645220e+06	3.757987e+07	2.949520e-07
4DPI_SAC_3_2	3.419064e+07	-4.386457e+07	2.949520e-07
4DPI_SAC_2_3	2.145412e+07	1.723141e+06	2.949520e-07
4DPI_SAC_2_2	-2.183335e+06	-1.551289e+07	2.949520e-07
4DPI_SAC_2_1	-1.893998e+07	-5.739152e+06	2.949520e-07
4DPI_SAC_1_3	-2.059662e+07	1.852929e+07	2.949520e-07
4DPI_SAC_1_2	3.062973e+06	-1.035837e+07	2.949520e-07
4DPI_SAC_1_1	-1.611010e+06	3.801341e+06	2.949520e-07
10DPI_interface_5_3	1.634717e+07	-9.292562e+04	2.949520e-07
10DPI_interface_4_3	-2.843180e+07	-7.313157e+06	2.949520e-07
10DPI_interface_4_2	7.446369e+06	-2.292092e+06	2.949520e-07
10DPI_interface_4_1	2.750186e+07	2.790485e+06	2.949520e-07
10DPI_interface_3_2	1.381678e+07	2.345572e+07	2.949520e-07
10DPI_interface_3_1	-3.987217e+07	-2.743997e+07	2.949520e-07
10DPI_cortex_4_4	-7.014139e+06	-1.328333e+07	2.949520e-07
10DPI_cortex_4_3	5.219814e+06	2.369728e+07	2.949520e-07
10DPI_cortex_3_3	6.556227e+06	-7.602979e+06	2.949520e-07
10DPI_cortex_3_2	-4.925120e+06	4.877634e+05	2.949520e-07
10DPI_SAC_4_3	8.374328e+05	-5.950123e+06	2.949520e-07
10DPI_SAC_4_2	-9.686167e+06	2.496185e+07	2.949520e-07
10DPI_SAC_4_1	3.640798e+06	-1.856121e+07	2.949520e-07
10DPI_SAC_3_3	-2.675078e+06	-1.110516e+07	2.949520e-07
10DPI_SAC_3_2	-2.930084e+06	-3.858284e+06	2.949520e-07

10DPI_SAC_3_1	2.071382e+07	1.198856e+07	2.949520e-07
10DPI_SAC_2_3	-7.880436e+06	-1.545762e+07	2.949520e-07
10DPI_SAC_2_2	-2.200700e+06	-1.017980e+07	2.949520e-07
10DPI_SAC_2_1	2.285816e+06	2.499442e+07	2.949520e-07

	Segment	K-means	PCA	TimePoint	BioRep	TechRep	\
index							
4DPI_interface_2_4		1		4DPI	2	4	
4DPI_interface_2_2		4		4DPI	2	2	
4DPI_interface_2_1		4		4DPI	2	1	
4DPI_interface_1_2		4		4DPI	1	2	
4DPI_interface_1_1		4		4DPI	1	1	
4DPI_cortex_3_3		0		4DPI	3	3	
4DPI_cortex_3_2		0		4DPI	3	2	
4DPI_cortex_2_3		0		4DPI	2	3	
4DPI_cortex_2_2		0		4DPI	2	2	
4DPI_cortex_2_1		4		4DPI	2	1	
4DPI_cortex_1_2		2		4DPI	1	2	
4DPI_cortex_1_1		2		4DPI	1	1	
4DPI_SAC_3_3		1		4DPI	3	3	
4DPI_SAC_3_2		1		4DPI	3	2	
4DPI_SAC_2_3		1		4DPI	2	3	
4DPI_SAC_2_2		1		4DPI	2	2	
4DPI_SAC_2_1		1		4DPI	2	1	
4DPI_SAC_1_3		1		4DPI	1	3	
4DPI_SAC_1_2		1		4DPI	1	2	
4DPI_SAC_1_1		1		4DPI	1	1	
10DPI_interface_5_3		3		10DPI	5	3	
10DPI_interface_4_3		3		10DPI	4	3	
10DPI_interface_4_2		3		10DPI	4	2	
10DPI_interface_4_1		3		10DPI	4	1	
10DPI_interface_3_2		1		10DPI	3	2	
10DPI_interface_3_1		3		10DPI	3	1	
10DPI_cortex_4_4		4		10DPI	4	4	
10DPI_cortex_4_3		0		10DPI	4	3	
10DPI_cortex_3_3		4		10DPI	3	3	
10DPI_cortex_3_2		0		10DPI	3	2	
10DPI_SAC_4_3		1		10DPI	4	3	
10DPI_SAC_4_2		1		10DPI	4	2	
10DPI_SAC_4_1		1		10DPI	4	1	
10DPI_SAC_3_3		3		10DPI	3	3	
10DPI_SAC_3_2		3		10DPI	3	2	
10DPI_SAC_3_1		3		10DPI	3	1	
10DPI_SAC_2_3		1		10DPI	2	3	
10DPI_SAC_2_2		1		10DPI	2	2	
10DPI_SAC_2_1		1		10DPI	2	1	

	Region	DPI_Region	BioRep_TechRep
index			
4DPI_interface_2_4	interface	4DPI_interface	2_4
4DPI_interface_2_2	interface	4DPI_interface	2_2
4DPI_interface_2_1	interface	4DPI_interface	2_1
4DPI_interface_1_2	interface	4DPI_interface	1_2
4DPI_interface_1_1	interface	4DPI_interface	1_1
4DPI_cortex_3_3	cortex	4DPI_cortex	3_3
4DPI_cortex_3_2	cortex	4DPI_cortex	3_2
4DPI_cortex_2_3	cortex	4DPI_cortex	2_3
4DPI_cortex_2_2	cortex	4DPI_cortex	2_2
4DPI_cortex_2_1	cortex	4DPI_cortex	2_1
4DPI_cortex_1_2	cortex	4DPI_cortex	1_2
4DPI_cortex_1_1	cortex	4DPI_cortex	1_1
4DPI_SAC_3_3	SAC	4DPI_SAC	3_3
4DPI_SAC_3_2	SAC	4DPI_SAC	3_2
4DPI_SAC_2_3	SAC	4DPI_SAC	2_3
4DPI_SAC_2_2	SAC	4DPI_SAC	2_2
4DPI_SAC_2_1	SAC	4DPI_SAC	2_1
4DPI_SAC_1_3	SAC	4DPI_SAC	1_3
4DPI_SAC_1_2	SAC	4DPI_SAC	1_2
4DPI_SAC_1_1	SAC	4DPI_SAC	1_1
10DPI_interface_5_3	interface	10DPI_interface	5_3
10DPI_interface_4_3	interface	10DPI_interface	4_3
10DPI_interface_4_2	interface	10DPI_interface	4_2
10DPI_interface_4_1	interface	10DPI_interface	4_1
10DPI_interface_3_2	interface	10DPI_interface	3_2
10DPI_interface_3_1	interface	10DPI_interface	3_1
10DPI_cortex_4_4	cortex	10DPI_cortex	4_4
10DPI_cortex_4_3	cortex	10DPI_cortex	4_3
10DPI_cortex_3_3	cortex	10DPI_cortex	3_3
10DPI_cortex_3_2	cortex	10DPI_cortex	3_2
10DPI_SAC_4_3	SAC	10DPI_SAC	4_3
10DPI_SAC_4_2	SAC	10DPI_SAC	4_2
10DPI_SAC_4_1	SAC	10DPI_SAC	4_1
10DPI_SAC_3_3	SAC	10DPI_SAC	3_3
10DPI_SAC_3_2	SAC	10DPI_SAC	3_2
10DPI_SAC_3_1	SAC	10DPI_SAC	3_1
10DPI_SAC_2_3	SAC	10DPI_SAC	2_3
10DPI_SAC_2_2	SAC	10DPI_SAC	2_2
10DPI_SAC_2_1	SAC	10DPI_SAC	2_1

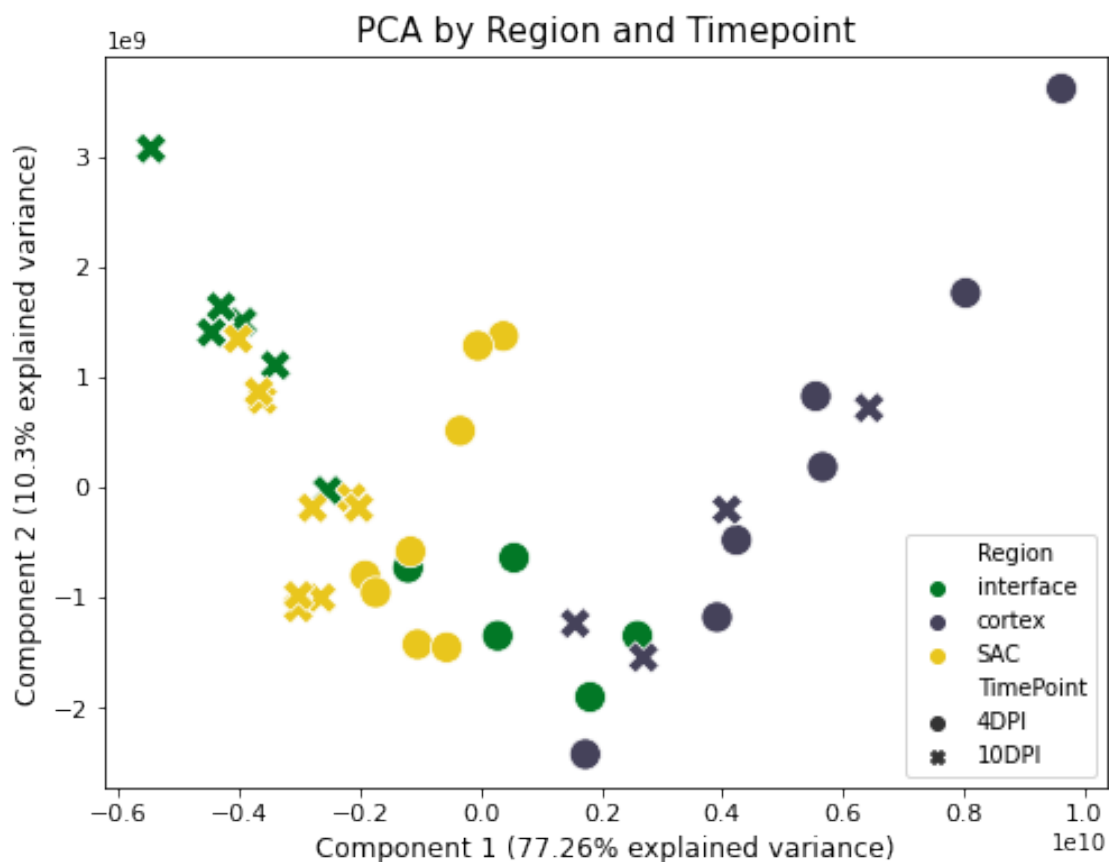
[39 rows x 3659 columns]

```
[49]: x_axis = df_segm_pca_kmeans['Component 1']
y_axis = df_segm_pca_kmeans['Component 2']
plt.figure(figsize=(8,6))
```

```

sns.scatterplot(x_axis, y_axis,
    →hue=df_segmpca_kmeans['Region'],palette=["#027826","#45425A","#E9C61D"],
    →style=df_segmpca_kmeans['TimePoint'],s=200,legend='full')
plt.title('PCA by Region and Timepoint', fontsize=15)
plt.tick_params(axis='both', which='major', labels=11)
plt.tick_params(axis='both', which='minor', labels=11)
plt.xlabel('Component 1 (' + str((pca_expl_var[0]*100).round(2)) + '% explained_
    →variance)', fontsize=12)
plt.ylabel('Component 2 (' + str((pca_expl_var[1]*100).round(2)) + '% explained_
    →variance)', fontsize=12)
plt.legend(loc='lower right')
plt.savefig('outs/' + date + '_combined_PCAwithRegionsandTimepoints.svg')

```



```

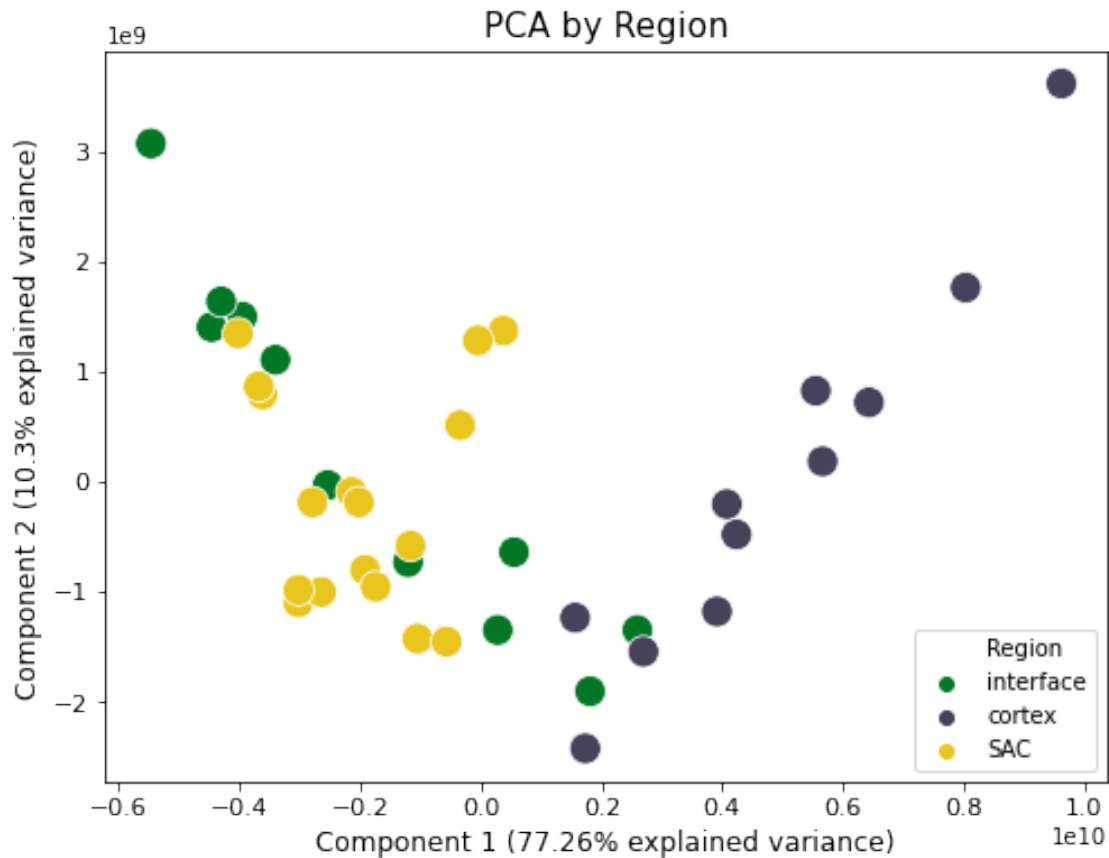
[50]: x_axis = df_segmpca_kmeans['Component 1']
y_axis = df_segmpca_kmeans['Component 2']
plt.figure(figsize=(8,6))
sns.scatterplot(x_axis, y_axis,
    →hue=df_segmpca_kmeans['Region'],palette=["#027826","#45425A","#E9C61D"],s=200,
    →legend='full')
plt.title('PCA by Region', fontsize=15)

```

```

plt.tick_params(axis='both', which='major', labels=11)
plt.tick_params(axis='both', which='minor', labels=11)
plt.xlabel('Component 1 (' + str((pca_expl_var[0]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.ylabel('Component 2 (' + str((pca_expl_var[1]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.legend(loc='lower right')
plt.savefig('outs/' + date + '_combined_PCAbbyRegion.svg')

```



```

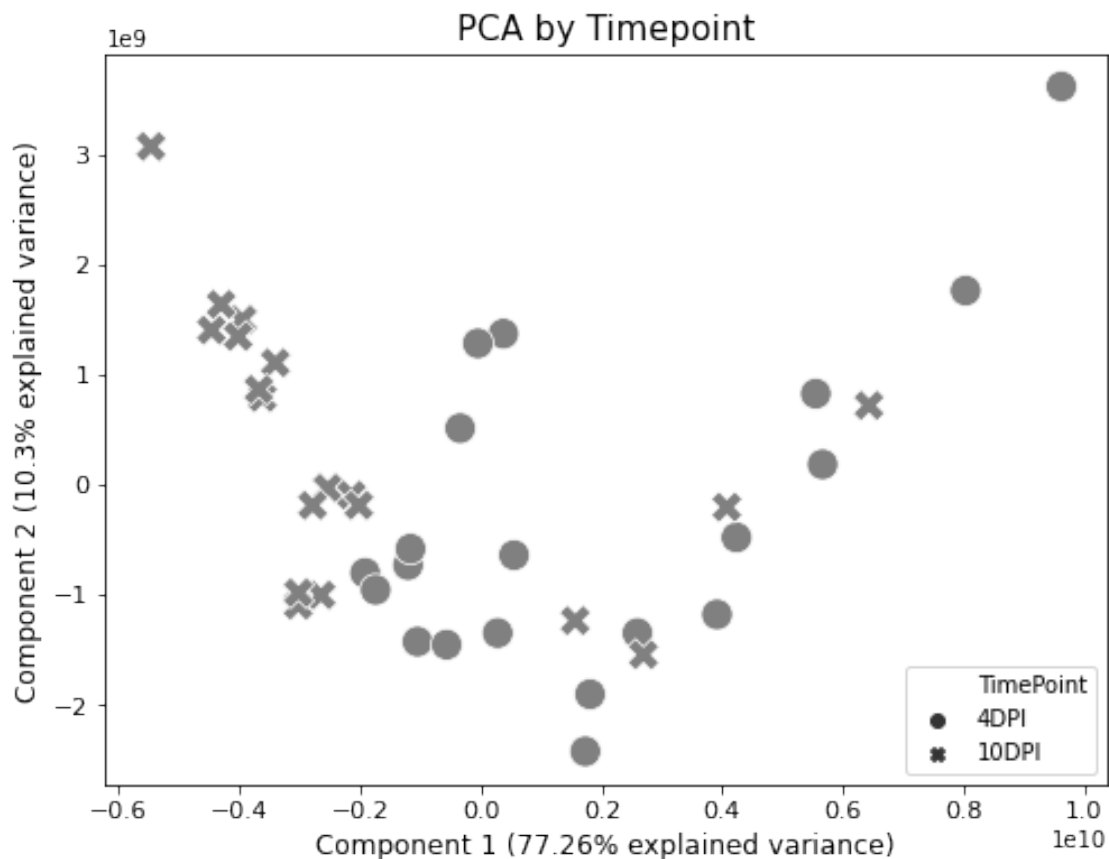
[51]: x_axis = df_segm_pca_kmeans['Component 1']
y_axis = df_segm_pca_kmeans['Component 2']
plt.figure(figsize=(8,6))
sns.scatterplot(x_axis,
                y_axis,
                color='grey',
                style=df_segm_pca_kmeans['TimePoint'],
                s=200,
                legend='full')
plt.title('PCA by Timepoint', fontsize=15)
plt.tick_params(axis='both', which='major', labels=11)

```

```

plt.tick_params(axis='both', which='minor', labelsiz=11)
plt.xlabel('Component 1 (' + str((pca_expl_var[0]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.ylabel('Component 2 (' + str((pca_expl_var[1]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.legend(loc='lower right')
plt.savefig('outs/' + date + '_combined_PCAbTimepoint.svg')

```



```

[54]: x_axis = df_segm_pca_kmeans['Component 1']
y_axis = df_segm_pca_kmeans['Component 2']
labels = df_segm_pca_kmeans['BioRep_TechRep']
plt.figure(figsize=(8,6))
sns.scatterplot(x_axis,
                y_axis,
                hue=df_segm_pca_kmeans['Region'],
                palette=["#027826", "#45425A", "#E9C61D"],
                s=60,
                legend='full')
plt.title('PCA by BioRep_TechRep', fontsize=15)
plt.tick_params(axis='both', which='major', labelsiz=11)

```

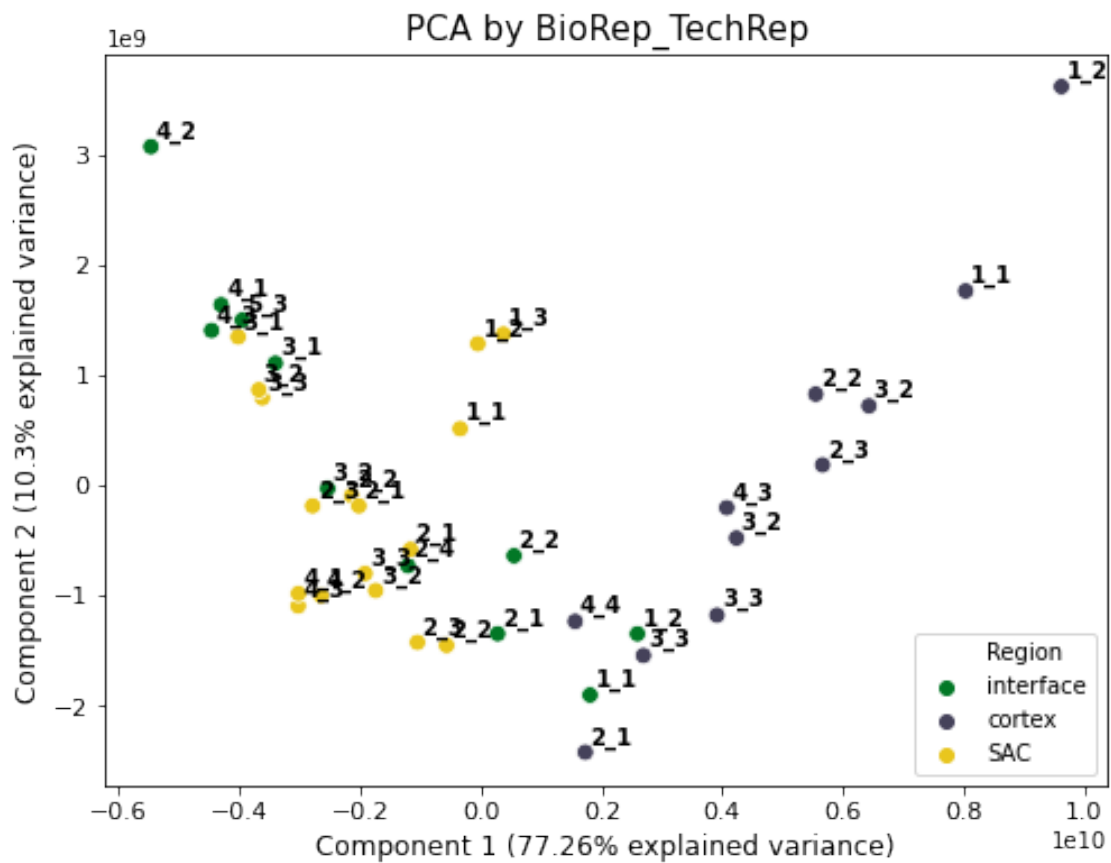
```

plt.tick_params(axis='both', which='minor', labelsz=11)
plt.xlabel('Component 1 (' + str((pca_expl_var[0]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.ylabel('Component 2 (' + str((pca_expl_var[1]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.legend(loc='lower right')

for line in range(0,df_segm_pca_kmeans.shape[0]):
    plt.annotate(xy = (x_axis[line], y_axis[line]),
                text = labels[line],
                horizontalalignment='left',
                verticalalignment='bottom',
                size='medium',
                color='black',
                weight='semibold',
                xytext=(2, 2),
                textcoords='offset points')

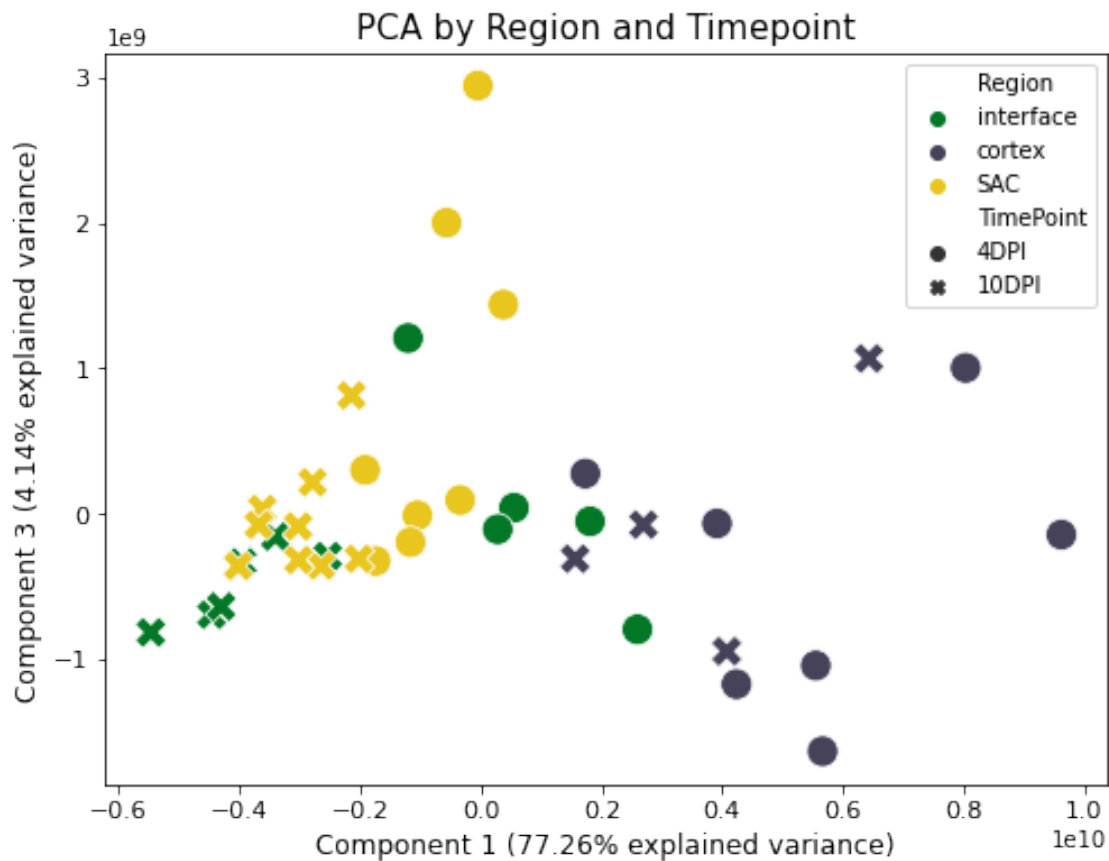
plt.savefig('outs/' + date + '_combined_PCABioRep_TechRep.svg')

```

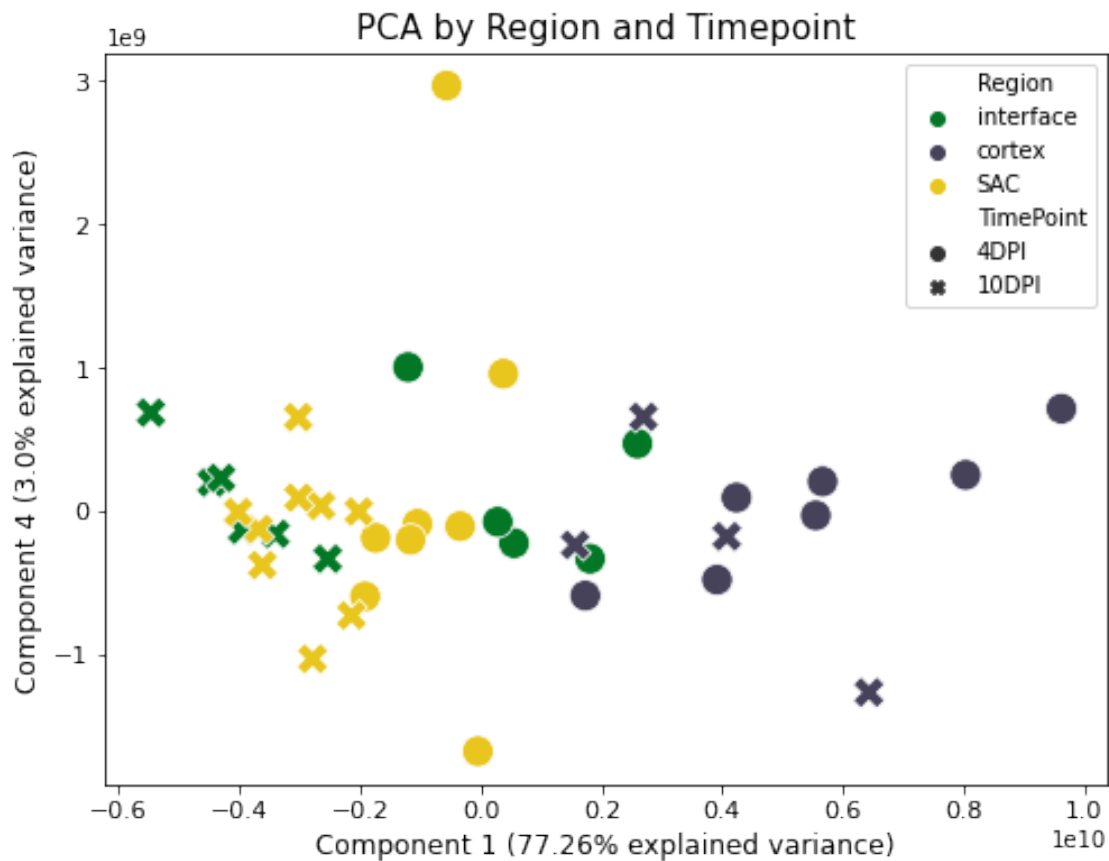




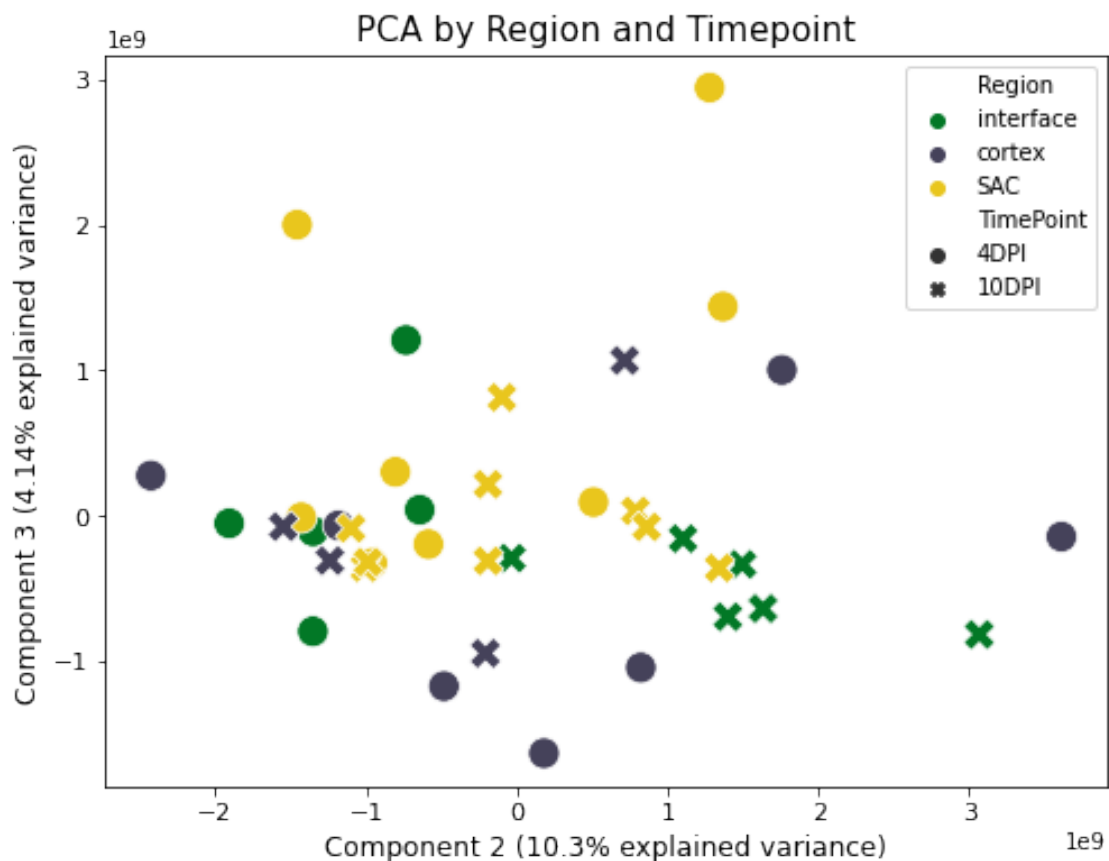
```
[55]: x_axis = df_segmpca_kmeans['Component 1']
y_axis = df_segmpca_kmeans['Component 3']
plt.figure(figsize=(8,6))
sns.scatterplot(x_axis,
                y_axis,
                hue=df_segmpca_kmeans['Region'],
                palette=["#027826", "#45425A", "#E9C61D"],
                style=df_segmpca_kmeans['TimePoint'],
                s=200,
                legend='full')
plt.title('PCA by Region and Timepoint', fontsize=15)
plt.tick_params(axis='both', which='major', labelsize=11)
plt.tick_params(axis='both', which='minor', labelsize=11)
plt.xlabel('Component 1 (' + str((pca_expl_var[0]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.ylabel('Component 3 (' + str((pca_expl_var[2]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.legend(loc='upper right')
plt.savefig('outs/' + date + '_combined_PCAwithRegionsandTimepoints_PC13.svg')
```



```
[56]: x_axis = df_segmpca_kmeans['Component 1']
y_axis = df_segmpca_kmeans['Component 4']
plt.figure(figsize=(8,6))
sns.scatterplot(x_axis,
                y_axis,
                hue=df_segmpca_kmeans['Region'],
                palette=["#027826", "#45425A", "#E9C61D"],
                style=df_segmpca_kmeans['TimePoint'],
                s=200,
                legend='full')
plt.title('PCA by Region and Timepoint', fontsize=15)
plt.tick_params(axis='both', which='major', labelsize=11)
plt.tick_params(axis='both', which='minor', labelsize=11)
plt.xlabel('Component 1 (' + str((pca_expl_var[0]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.ylabel('Component 4 (' + str((pca_expl_var[3]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.legend(loc='upper right')
plt.savefig('outs/' + date + '_combined_PCawithRegionsandTimepoints_PC14.svg')
```



```
[57]: x_axis = df_segmpca_kmeans['Component 2']
y_axis = df_segmpca_kmeans['Component 3']
plt.figure(figsize=(8,6))
sns.scatterplot(x_axis, y_axis,
                hue=df_segmpca_kmeans['Region'],
                palette=["#027826", "#45425A", "#E9C61D"],
                style=df_segmpca_kmeans['TimePoint'],
                s=200,
                legend='full')
plt.title('PCA by Region and Timepoint', fontsize=15)
plt.tick_params(axis='both', which='major', labels=11)
plt.tick_params(axis='both', which='minor', labels=11)
plt.xlabel('Component 2 (' + str((pca_expl_var[1]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.ylabel('Component 3 (' + str((pca_expl_var[2]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.legend(loc='upper right')
plt.savefig('outs/' + date + '_combined_PCAwithRegionsandTimepoints_PC23.svg')
```



```
[58]: palette_kmeans = ["#dd7373", "#51a3a3", "#37d4fb", "#0d3b66", "#7b3e19"]
```

```
[59]: x_axis = df_segm_pca_kmeans['Component 1']
y_axis = df_segm_pca_kmeans['Component 2']
plt.figure(figsize=(8,6))

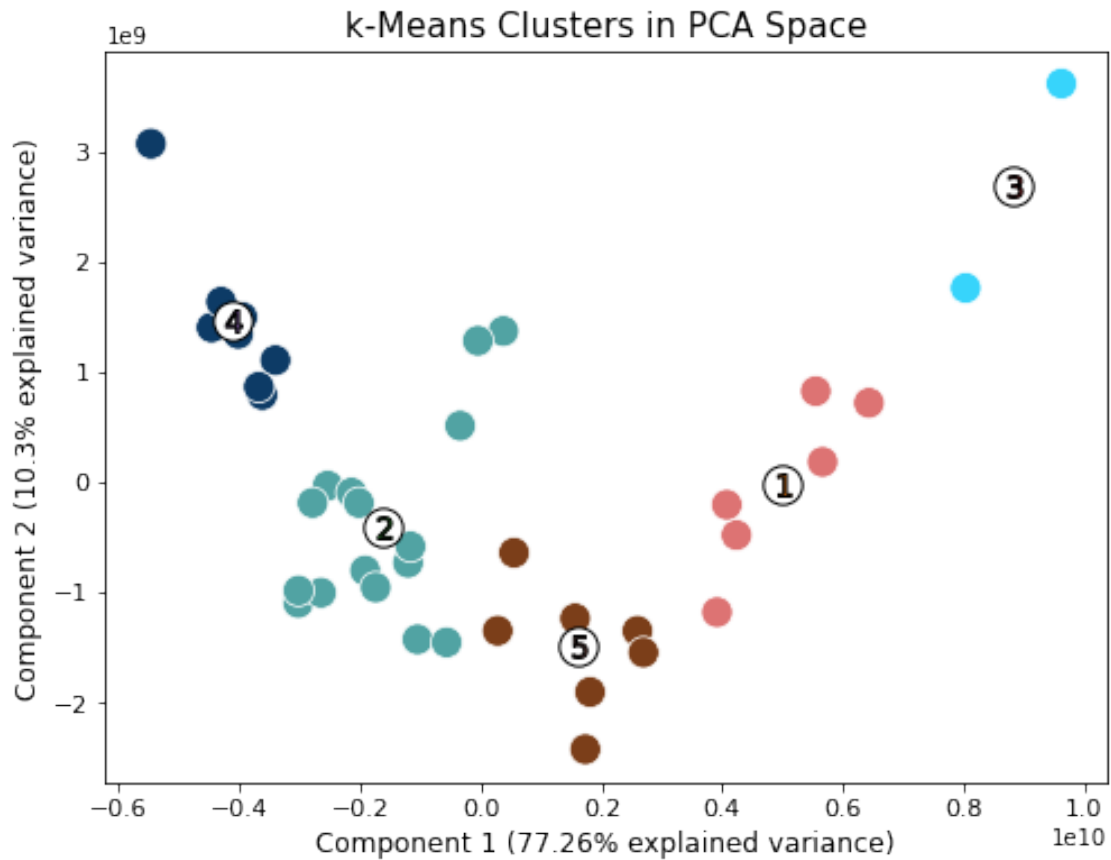
sns.scatterplot(x_axis, y_axis, hue=df_segm_pca_kmeans['Segment K-means_
→PCA'],palette=palette_kmeans,s=200, legend=None)

plt.title('k-Means Clusters in PCA Space', fontsize=15)
plt.tick_params(axis='both', which='major', labelsize=11)
plt.tick_params(axis='both', which='minor', labelsize=11)
plt.xlabel('Component 1 (' + str((pca_expl_var[0]*100).round(2)) + '% explained_
→variance)', fontsize=12)
plt.ylabel('Component 2 (' + str((pca_expl_var[1]*100).round(2)) + '% explained_
→variance)', fontsize=12)

plt.scatter(cluster_centers[:, 0], cluster_centers[:, 1], marker='o',
            c="white", alpha=1, s=300, edgecolor='k')

for i, c in enumerate(cluster_centers):
    i = i+1
    plt.scatter(c[0], c[1], marker='$_d$' % i, alpha=1,
                s=100, edgecolor='k')

plt.savefig('outs/' + date + '_combined_kClustersinPCAspace.svg')
```



```
[60]: print(cluster_centers)
      print(type(cluster_centers))
```

```
[[ 4.97631614e+09 -2.62226294e+07 -6.36404452e+08 -2.77132768e+08
 -9.25103083e+07  3.85743372e+07  2.36238066e+07  1.51952879e+08
 -1.85638030e+08  3.75403214e+06 -3.76877118e+07  4.74203082e+07
 -3.99385925e+07 -1.51797031e+07  9.25132806e+06 -3.72809384e+05
 -2.22019011e+07  1.11803278e+07 -2.58436706e+07 -2.52654634e+07
 -1.57854945e+07  2.28581287e+06  6.09748976e+06 -3.96590426e+06
 -2.99693052e+06  1.38742397e+07  2.62176888e+06 -1.29283084e+07
  9.97017102e+06 -1.04632069e+07  4.54286107e+06 -1.08787849e+07
 -6.15299352e+06  5.28378862e+06  5.87929981e+06  3.46793520e+06
 -4.14142833e+06  6.79536245e+06  2.94952004e-07]
[-1.61461434e+09 -4.01790671e+08  4.42629517e+08  4.61323893e+07
 1.85518933e+08 -1.18867632e+08 -6.88108370e+07  2.56672292e+05
 6.36768016e+06 -2.46156286e+07  3.84413546e+07 -6.18636555e+06
 -2.09401395e+07 -4.27549157e+07 -4.49396285e+06 -1.24708782e+07
 1.52469667e+07  1.92247824e+07 -1.34558012e+07 -3.10130684e+06
 8.44914232e+06  1.24654478e+07 -7.24884020e+06  1.98426488e+06
 3.98094322e+06 -4.26226131e+06  7.01575538e+06 -2.77713256e+06]
```

```

-4.79593764e+06 -4.11294241e+06  1.05162503e+06 -4.18349027e+06
 1.77869719e+06 -2.19594124e+06 -1.73424287e+06 -2.94816893e+06
 5.48309415e+04  2.95050822e+06  2.94952004e-07]
[ 8.82376536e+09  2.68506889e+09  4.27130564e+08  4.81770218e+08
 4.83561663e+08 -5.43207091e+07  1.66048357e+07 -1.52276571e+08
 3.08675403e+08  4.16169643e+07  2.99056274e+07 -3.11807489e+07
 7.14139984e+06  1.01589270e+07 -6.83929204e+07 -1.64718554e+07
 1.45543071e+07  8.63889376e+06  6.69433340e+07 -1.73647886e+07
 4.07586421e+07  2.33542144e+07  1.13921544e+07  7.87203352e+06
-8.97216021e+05 -8.21139372e+06  2.05715124e+07  1.56714453e+07
-3.72400508e+06  2.08730651e+07 -2.62418375e+06  3.21957118e+05
 1.41653870e+07  2.89186214e+06 -6.73519836e+06  2.10680503e+06
 1.60597906e+07 -1.33788951e+07  2.94952004e-07]
[-4.11152905e+09  1.46079621e+09 -3.84124687e+08  3.78848633e+07
-1.50634152e+08  2.13972599e+08  8.58592018e+07  3.08719807e+06
-4.21398906e+07  2.21866173e+07 -3.56259276e+07  8.99383171e+06
 2.71810883e+07  4.42505564e+07  9.87614677e+06 -8.19533924e+06
-1.00648921e+07 -1.87364589e+07  1.39875084e+07  8.16244165e+05
 1.56140635e+06 -1.00535558e+07  8.41426936e+06 -3.98084782e+06
-1.01625394e+07  7.38596103e+06 -2.22751894e+06  6.20357815e+06
 5.74169585e+06  5.20445251e+06 -3.76863355e+05  3.30182448e+06
-1.77114483e+06  6.38039515e+06  2.68059014e+06  5.08173281e+06
-2.37488472e+05 -4.66531742e+06  2.94952004e-07]
[ 1.60294776e+09 -1.49578871e+09 -1.49272741e+08 -4.88487086e+07
-3.10755883e+08  9.61381523e+06  3.41638953e+07 -9.08526389e+07
 1.04530517e+08  1.57998568e+07 -2.33913196e+07 -2.78755938e+07
 4.89917544e+07  5.72620805e+07  1.05960148e+07  4.28967617e+07
-8.47564835e+06 -3.45806573e+07  1.77954438e+07  3.27733306e+07
-1.92116922e+07 -2.56345749e+07 -1.52885128e+06  1.16427184e+06
 5.34017695e+06 -8.24487972e+06 -2.16150818e+07  5.86177940e+06
-3.08165437e+06  6.45779565e+06 -5.11712751e+06  1.50214348e+07
-8.14687048e+05 -7.62779391e+06 -2.21460533e+06 -2.64348297e+06
-8.92628359e+05 -3.41428238e+06  2.94952004e-07]]
<class 'numpy.ndarray'>

```

```
[61]: cluster_centers.shape
```

```
[61]: (5, 39)
```

```
[62]: components.shape
```

```
[62]: (39, 3613)
```

```
[63]: centers = cluster_centers.dot(components)
      centers.shape
```

```
[63]: (5, 3613)
```

```
[64]: centers = pd.DataFrame(centers,
```

```

        index=['Cluster 1', 'Cluster 2', 'Cluster 3', 'Cluster_
→4', 'Cluster 5'],
        columns = X.columns)
centers.to_csv('outs/' + date + '_combined_centers_annotated.csv')

```

```

[65]: centers_t = centers.T
      centers_t.shape

```

```

[65]: (3613, 5)

```

```

[66]: pGroups_proc.shape

```

```

[66]: (3613, 371)

```

```

[67]: #plot only staph markers
      #first pull only staph index then pull centers_t based on that index

staph_idx = pGroups_proc.index[pGroups_proc['Fasta headers'].str.
→contains('aureus', regex=False)].tolist()
centers_t_staph = centers_t.iloc[staph_idx]
centers_t_staph.index = centers_t_staph.index.str.split(';',0).str[0]
centers_t_staph.to_csv('outs/' + date + '_staph_centers_annotated.csv')
centers_t_staph

```

```

[67]:
      Cluster 1      Cluster 2      Cluster 3      Cluster 4 \
Protein IDs
Q2FER6      -3.003200e+05  1.624625e+05 -3.003200e+05  2.381750e+05
Q2FEQ8      -8.424872e+03 -8.424872e+03 -8.424872e+03  3.264638e+04
Q2FHT6      -4.496051e+04  1.489261e+04 -4.496051e+04  5.451574e+04
Q2FGF0      -2.179487e+04 -2.179487e+04 -2.179487e+04  8.445513e+04
Q2FJA2      -9.495641e+03 -9.495641e+03 -9.495641e+03  3.679561e+04
...
A0A2C9TKS1  -2.888718e+04 -2.888718e+04 -2.888718e+04  1.119378e+05
A0A1S5YLT4  -2.152005e+06  1.383074e+05  2.077899e+07 -1.974355e+06
A0A2C9TU38  -8.591956e+07  3.073015e+07 -8.591956e+07  9.707224e+07
W8UAB2      -1.310485e+05 -2.405221e+04 -1.310485e+05  2.938203e+05
Q2FH01      -5.815923e+04 -5.815923e+04 -5.815923e+04  2.253670e+05

```

```

      Cluster 5
Protein IDs
Q2FER6      -3.003200e+05
Q2FEQ8      -8.424872e+03
Q2FHT6      -4.496051e+04
Q2FGF0      -2.179487e+04
Q2FJA2      -9.495641e+03
...
A0A2C9TKS1  -2.888718e+04
A0A1S5YLT4  -2.152005e+06
A0A2C9TU38  -8.298626e+07
W8UAB2      -1.310485e+05

```

Q2FH01            -5.815923e+04

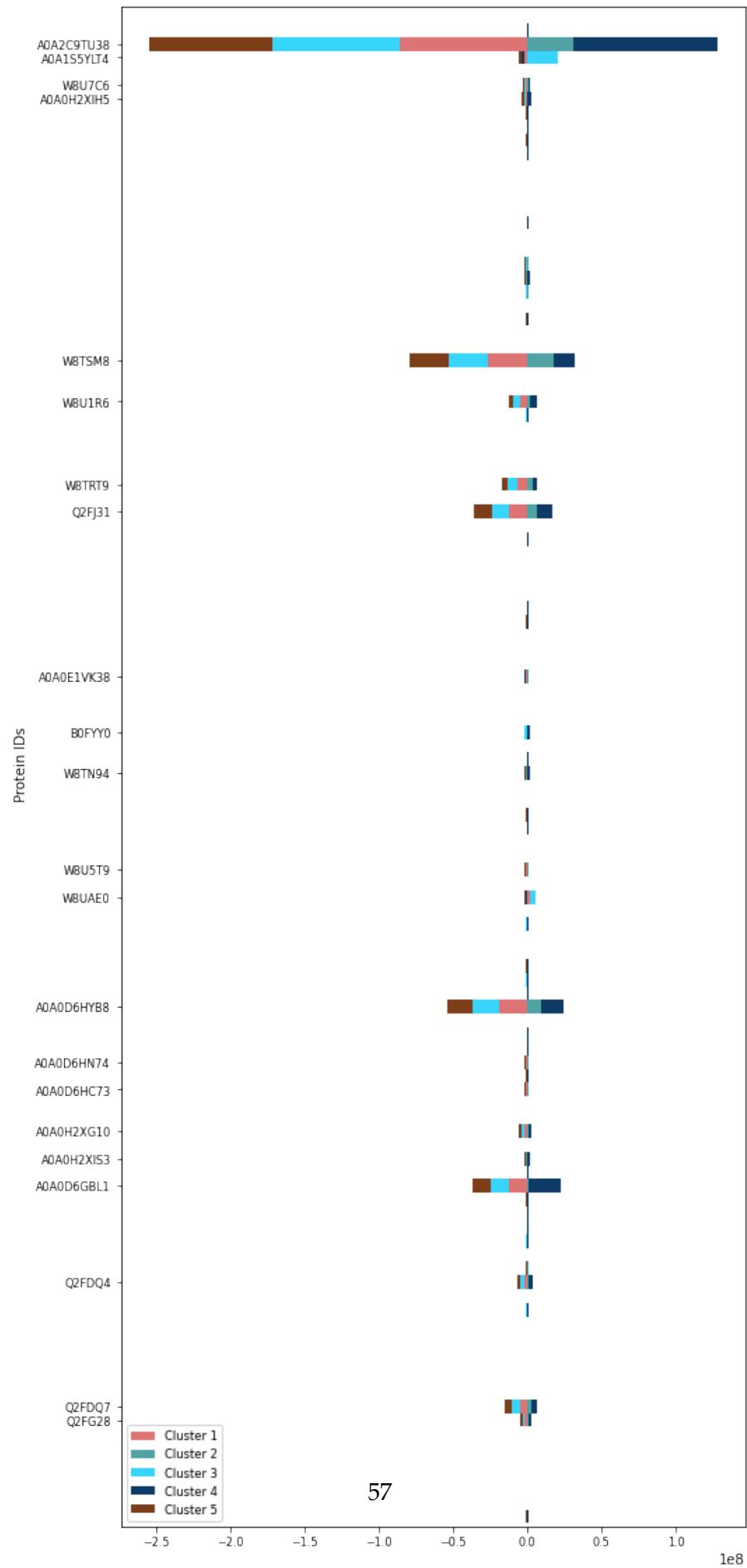
[110 rows x 5 columns]

```
[68]: summed = centers_t_staph.abs().sum(axis=1)
sorted_mylist = sorted(((v, i) for i, v in enumerate(summed)), reverse=True)[0:
    ↳22]
final = [lis[1] for lis in sorted_mylist]
finalID = centers_t_staph.iloc[final].index

centers_t_staph.iloc[final].to_csv('outs/' + date + '_finalID_staph_all.csv')

centers_t_staph.plot(kind="barh", stacked=True, width=1, figsize = (8,20),
    ↳color=palette_kmeans)
plt.tick_params(axis='both', which='major', labelsize=8)
plt.yticks(final, finalID)
#plt.xlim(-0.5e9, 0.5e9)
plt.legend(fontsize='small')
plt.savefig('outs/' + date + '_staph_kmeansCentroids_all.svg')
```





```
[69]: #plot only mouse markers
#first pull only mouse index then pull centers_t based on that index

mouse_idx = pGroups_proc.index[pGroups_proc['Fasta headers'].str.
    ↳contains('musculus', regex=False, na=False)].tolist()
centers_t_mouse = centers_t.iloc[mouse_idx]
centers_t_mouse
centers_t_mouse.index = centers_t_mouse.index.str.split('; ',0).str[0]
centers_t_mouse.to_csv('outs/' + date + '_mouse_centers_annotated.csv')
centers_t_mouse
```

```
[69]:
```

	Cluster 1	Cluster 2	Cluster 3	Cluster 4 \
Protein IDs				
G3UZW7	-4.494667e+04	50722.708333	-44946.666664	-17170.416668
O70589	4.249538e+04	-15230.865385	-29954.615385	-29954.615385
AOA140T8M9	-1.959910e+04	63072.564102	-102757.435898	-102757.435897
AOA075B5M7	-2.005736e+04	10793.891024	-20057.358967	16019.766022
AOA075B5P3	-2.679115e+05	30169.086538	-267911.538460	441995.961538
...	...	...	...	...
Q9Z2X1	-2.716608e+06	941850.641026	-180424.358974	517575.641026
Q9Z2Z6	3.939454e+05	-255760.032051	429668.717949	-327846.282051
Q9Z315	-5.958347e+05	189973.381410	-820199.743590	427097.756410
S4R2J6	9.806026e+03	5933.317308	614597.692308	-84808.557692
S4R192	-9.051179e+04	1393.205128	-90511.794872	166923.205128

	Cluster 5
Protein IDs	
G3UZW7	-44946.666667
O70589	41181.098901
AOA140T8M9	19429.706960
AOA075B5M7	-20057.358973
AOA075B5P3	-267911.538462
...	...
Q9Z2X1	-364245.787546
Q9Z2Z6	498845.860806
Q9Z315	-177278.315018
S4R2J6	-100642.307692
S4R192	-90511.794872

[3503 rows x 5 columns]

```
[71]: #plot centroids with top 1% proteins labeled

summed = centers_t_mouse.abs().sum(axis=1)
sorted_mylist = sorted(((v, i) for i, v in enumerate(summed)), reverse=True)[0:
    ↳35]
```

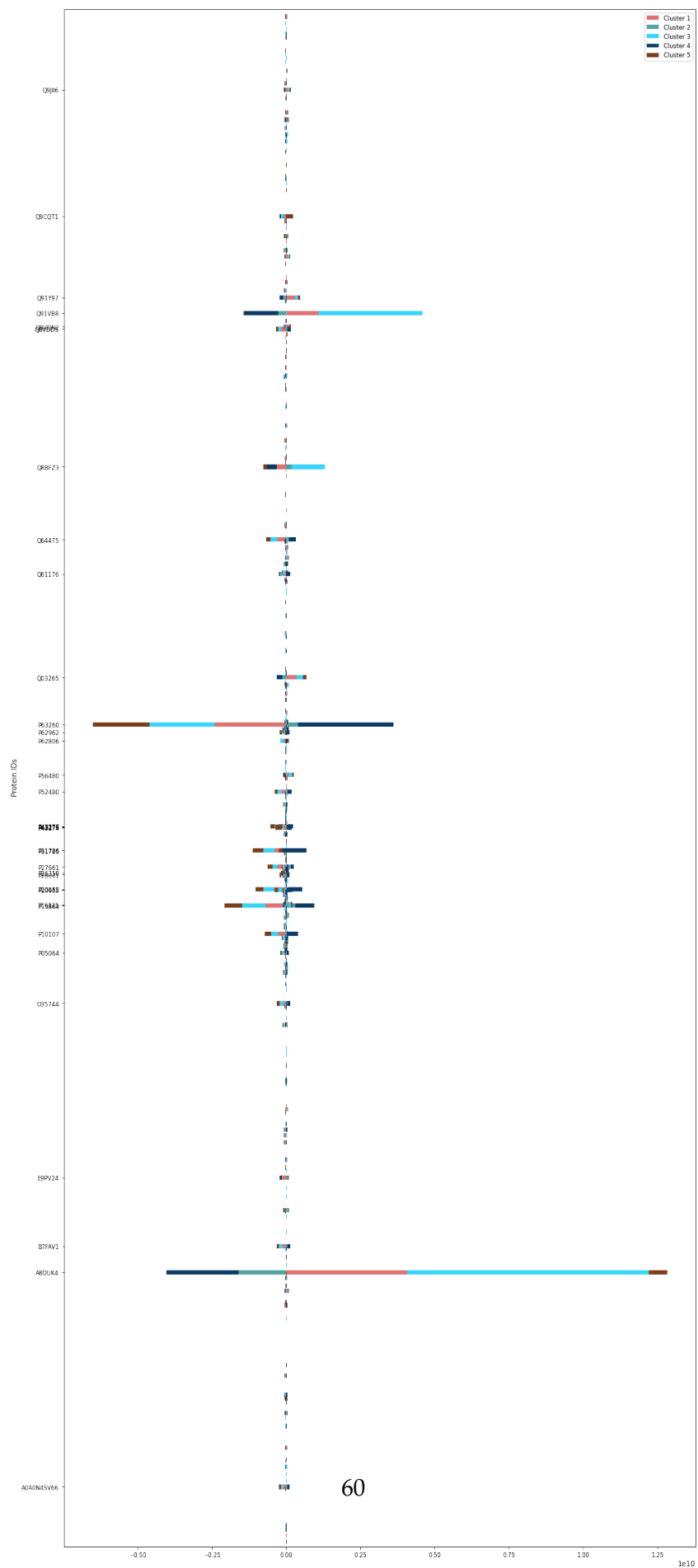
```

final = [lis[1] for lis in sorted_mylist]
finalID = centers_t_mouse.iloc[final].index

centers_t_mouse.to_csv('outs/' + date + '_finalID_mouse_all.csv')

centers_t_mouse.plot(kind="barh", stacked=True, width=10, figsize = (16,40),
    →color=palette_kmeans)
plt.tick_params(axis='both', which='major', labelsize=8)
plt.yticks(final, finalID)
#plt.xlim(-3e11, 3e11)
plt.legend(fontsize='small')
plt.savefig('outs/' + date + '_mouse_kmeansCentroids_all_top35.svg')

```



## 0.4 Gene Ontology Enrichment Scores

### 0.4.1 Extract top 100 proteins driving the clustering and run gene ontology enrichment analysis using their LFQ intensity values after normalization

```
[75]: allProt = pd.read_csv('outs/' + date + '_finalID_mouse_all.csv', index_col = 0)
allProt.shape
```

```
[75]: (3503, 5)
```

```
[73]: summed = allProt.abs().sum(axis=1)
sorted_mylist = sorted(((v, i) for i, v in enumerate(summed)), reverse=True)[0:
    ↪175]
final = [lis[1] for lis in sorted_mylist]
finalID = allProt.iloc[final].index
finalID
```

```
[73]: Index(['A8DUK4', 'P63260', 'Q91VB8', 'P15864', 'Q8BFZ3', 'P31725', 'P20152',
    'P10107', 'Q64475', 'Q03265',
    ...,
    'Q3UZ39', 'Q8K2B3', 'P54071', 'Q5EBP8', 'Q3UHL6', 'P22599', 'Q64523',
    'F8WIT2', 'Q3THW5', 'P00920'],
    dtype='object', name=' Protein IDs', length=175)
```

```
[77]: lfq = pd.read_csv('outs/' + date + '_combined_lfq.csv', index_col=0)
lfq = lfq.transpose()
lfq = lfq.drop(['TimePoint', 'TechRep', 'BioRep', 'Region',
    ↪'DPI_Region_BioRep', 'DPI_Region', 'BioRep_TechRep'], axis=0)
lfq
```

```
[77]: index                                4DPI_interface_2_4 \
G3UZW7;A0A023T778;Q9CQL1;P61327                                0
O70589;A0A067XG53;F6Y9I5                                        0
A0A140T8M9;A0A140T8M0;A0A0B4J1I0;A0A075B5N0;A0A...            0
A0A075B5M7;A0A0G2JDV4;A0A0B4J1J2;A0A0B4J1J1                0
A0A075B5P3;A0A0A6YVP0;P01867                                    0
...                                                              ...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2                    5.246e+06
Q9Z2Z6                                                         0
Q9Z315;A0A494B9E9                                              954220
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5                    0
S4R192                                                         0

index                                4DPI_interface_2_2 \
G3UZW7;A0A023T778;Q9CQL1;P61327                                0
O70589;A0A067XG53;F6Y9I5                                        0
A0A140T8M9;A0A140T8M0;A0A0B4J1I0;A0A075B5N0;A0A...            0
```

AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	3.0364e+06
Q9Z2Z6	668000
Q9Z315;AOA494B9E9	1.3774e+06
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	0
index	4DPI_interface_2_1 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	3.9714e+06
Q9Z2Z6	0
Q9Z315;AOA494B9E9	922440
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	0
index	4DPI_interface_1_2 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	9.2853e+06
Q9Z2Z6	1.4414e+06
Q9Z315;AOA494B9E9	0
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	0
index	4DPI_interface_1_1 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	6.0399e+06
Q9Z2Z6	1.1892e+06
Q9Z315;AOA494B9E9	1.1176e+06
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	0

index	4DPI_cortex_3_3 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	2.6449e+06
Q9Z2Z6	1.3931e+06
Q9Z315;AOA494B9E9	638210
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	0

index	4DPI_cortex_3_2 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	0
Q9Z2Z6	950880
Q9Z315;AOA494B9E9	661230
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	0

index	4DPI_cortex_2_3 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	2.4288e+06
Q9Z2Z6	1.0941e+06
Q9Z315;AOA494B9E9	0
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	662690
S4R192	0

index	4DPI_cortex_2_2 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;P01867	0
...	...

Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	2.6804e+06	
Q9Z2Z6	1.1699e+06	
Q9Z315;AOA494B9E9	494860	
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	
S4R192	0	
index	4DPI_cortex_2_1	\
G3UZW7;AOA023T778;Q9CQL1;P61327	0	...
O70589;AOA067XG53;F6Y9I5	497950	...
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0	...
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0	...
AOA075B5P3;AOA0A6YVP0;P01867	0	...
...	...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	2.0776e+06	...
Q9Z2Z6	1.1146e+06	...
Q9Z315;AOA494B9E9	674540	...
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	...
S4R192	0	...
index	10DPI_cortex_3_2	\
G3UZW7;AOA023T778;Q9CQL1;P61327	0	
O70589;AOA067XG53;F6Y9I5	434700	
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	498950	
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0	
AOA075B5P3;AOA0A6YVP0;P01867	0	
...	...	
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	2.1112e+06	
Q9Z2Z6	0	
Q9Z315;AOA494B9E9	556170	
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	
S4R192	0	
index	10DPI_SAC_4_3	\
G3UZW7;AOA023T778;Q9CQL1;P61327	0	
O70589;AOA067XG53;F6Y9I5	0	
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	1.6741e+06	
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	202930	
AOA075B5P3;AOA0A6YVP0;P01867	1.2076e+06	
...	...	
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	6.5735e+06	
Q9Z2Z6	338860	
Q9Z315;AOA494B9E9	1.1378e+06	
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	
S4R192	0	
index	10DPI_SAC_4_2	\
G3UZW7;AOA023T778;Q9CQL1;P61327	0	



O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	431080
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	290690
AOA075B5P3;AOA0A6YVP0;PO1867	1.5261e+06
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	6.941e+06
Q9Z2Z6	0
Q9Z315;AOA494B9E9	1.3022e+06
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	0
index	10DPI_SAC_4_1 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	548100
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;PO1867	1.1656e+06
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	6.2944e+06
Q9Z2Z6	0
Q9Z315;AOA494B9E9	0
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	326950
index	10DPI_SAC_3_3 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;PO1867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	4.6115e+06
Q9Z2Z6	0
Q9Z315;AOA494B9E9	1.7312e+06
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0
S4R192	423330
index	10DPI_SAC_3_2 \
G3UZW7;AOA023T778;Q9CQL1;P61327	0
O70589;AOA067XG53;F6Y9I5	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0
AOA075B5P3;AOA0A6YVP0;PO1867	0
...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	6.5767e+06
Q9Z2Z6	0
Q9Z315;AOA494B9E9	1.4447e+06

S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	
S4R192	0	
index	10DPI_SAC_3_1	\
G3UZW7;AOA023T778;Q9CQL1;P61327	0	
O70589;AOA067XG53;F6Y9I5	0	
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0	
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0	
AOA075B5P3;AOA0A6YVP0;P01867	724260	
...	...	
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	5.9742e+06	
Q9Z2Z6	0	
Q9Z315;AOA494B9E9	1.0595e+06	
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	
S4R192	0	
index	10DPI_SAC_2_3	\
G3UZW7;AOA023T778;Q9CQL1;P61327	0	
O70589;AOA067XG53;F6Y9I5	0	
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0	
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0	
AOA075B5P3;AOA0A6YVP0;P01867	0	
...	...	
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	2.8952e+06	
Q9Z2Z6	0	
Q9Z315;AOA494B9E9	3.3433e+06	
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	
S4R192	0	
index	10DPI_SAC_2_2	10DPI_SAC_2_1
G3UZW7;AOA023T778;Q9CQL1;P61327	0	0
O70589;AOA067XG53;F6Y9I5	0	0
AOA140T8M9;AOA140T8M0;AOA0B4J1I0;AOA075B5N0;AOA...	0	0
AOA075B5M7;AOA0G2JDV4;AOA0B4J1J2;AOA0B4J1J1	0	0
AOA075B5P3;AOA0A6YVP0;P01867	0	0
...	...	...
Q9Z2X1;J3QMT0;J3QM80;J3QP45;J3QMV8;J3QNH2	3.8473e+06	6.8296e+06
Q9Z2Z6	0	0
Q9Z315;AOA494B9E9	800330	2.3382e+06
S4R2J6;S4R165;S4R1X7;S4R2C1;S4R187;S4R2F5	0	0
S4R192	0	0

[3613 rows x 39 columns]

```
[78]: lfq.index = lfq.index.str.split(';',0).str[0]
lfq
```

```

[78]: index      4DPI_interface_2_4 4DPI_interface_2_2 4DPI_interface_2_1 \
G3UZW7              0              0              0
070589              0              0              0
AOA140T8M9          0              0              0
AOA075B5M7          0              0              0
AOA075B5P3          0              0              0
...                ...                ...                ...
Q9Z2X1              5.246e+06          3.0364e+06          3.9714e+06
Q9Z2Z6              0              668000              0
Q9Z315              954220          1.3774e+06          922440
S4R2J6              0              0              0
S4R192              0              0              0

index      4DPI_interface_1_2 4DPI_interface_1_1 4DPI_cortex_3_3 \
G3UZW7              0              0              0
070589              0              0              0
AOA140T8M9          0              0              0
AOA075B5M7          0              0              0
AOA075B5P3          0              0              0
...                ...                ...                ...
Q9Z2X1              9.2853e+06          6.0399e+06          2.6449e+06
Q9Z2Z6              1.4414e+06          1.1892e+06          1.3931e+06
Q9Z315              0              1.1176e+06          638210
S4R2J6              0              0              0
S4R192              0              0              0

index      4DPI_cortex_3_2 4DPI_cortex_2_3 4DPI_cortex_2_2 4DPI_cortex_2_1 \
G3UZW7              0              0              0              0
070589              0              0              0          497950
AOA140T8M9          0              0              0              0
AOA075B5M7          0              0              0              0
AOA075B5P3          0              0              0              0
...                ...                ...                ...                ...
Q9Z2X1              0              2.4288e+06          2.6804e+06          2.0776e+06
Q9Z2Z6              950880          1.0941e+06          1.1699e+06          1.1146e+06
Q9Z315              661230              0          494860          674540
S4R2J6              0              662690              0              0
S4R192              0              0              0              0

index      ... 10DPI_cortex_3_2 10DPI_SAC_4_3 10DPI_SAC_4_2 10DPI_SAC_4_1 \
G3UZW7      ...              0              0              0              0
070589      ...              434700              0              0              0
AOA140T8M9  ...              498950          1.6741e+06          431080          548100
AOA075B5M7  ...              0              202930          290690              0
AOA075B5P3  ...              0              1.2076e+06          1.5261e+06          1.1656e+06
...          ...                ...                ...                ...                ...
Q9Z2X1      ...              2.1112e+06          6.5735e+06          6.941e+06          6.2944e+06

```

Q9Z2Z6	...	0	338860	0	0
Q9Z315	...	556170	1.1378e+06	1.3022e+06	0
S4R2J6	...	0	0	0	0
S4R192	...	0	0	0	326950

index	10DPI_SAC_3_3	10DPI_SAC_3_2	10DPI_SAC_3_1	10DPI_SAC_2_3	\
G3UZW7	0	0	0	0	
O70589	0	0	0	0	
AOA140T8M9	0	0	0	0	
AOA075B5M7	0	0	0	0	
AOA075B5P3	0	0	724260	0	
...	...	...	...	...	
Q9Z2X1	4.6115e+06	6.5767e+06	5.9742e+06	2.8952e+06	
Q9Z2Z6	0	0	0	0	
Q9Z315	1.7312e+06	1.4447e+06	1.0595e+06	3.3433e+06	
S4R2J6	0	0	0	0	
S4R192	423330	0	0	0	

index	10DPI_SAC_2_2	10DPI_SAC_2_1
G3UZW7	0	0
O70589	0	0
AOA140T8M9	0	0
AOA075B5M7	0	0
AOA075B5P3	0	0
...	...	...
Q9Z2X1	3.8473e+06	6.8296e+06
Q9Z2Z6	0	0
Q9Z315	800330	2.3382e+06
S4R2J6	0	0
S4R192	0	0

[3613 rows x 39 columns]

```
[79]: df = lfq.loc[finalID]
df
```

```
[79]: index      4DPI_interface_2_4 4DPI_interface_2_2 4DPI_interface_2_1 \
Protein IDs
A8DUK4      1.7442e+09      3.4139e+09      2.9423e+09
P63260      4.1307e+09      3.378e+09      2.9545e+09
Q91VB8      1.7297e+09      1.9171e+09      1.4639e+09
P15864      9.1421e+08      6.0707e+08      3.477e+08
Q8BFZ3      1.6431e+09      500220      6.6487e+06
...
P22599      2.9736e+07      2.3138e+07      3.2037e+07
Q64523      4.6711e+07      8.0783e+07      3.532e+07
F8WIT2      2.2663e+07      1.4908e+07      3.6708e+07
Q3THW5      3.9525e+07      3.0173e+07      3.4892e+07
```

P00920	2.3748e+07	5.1147e+06	2.2897e+07
--------	------------	------------	------------

index	4DPI_interface_1_2	4DPI_interface_1_1	4DPI_cortex_3_3	\
Protein IDs				
A8DUK4	5.1737e+09	3.8777e+09	5.782e+09	
P63260	1.771e+09	1.6077e+09	1.2741e+09	
Q91VB8	1.3123e+09	1.8905e+09	2.7786e+09	
P15864	4.7936e+08	5.7903e+08	2.5711e+08	
Q8BFZ3	4.4522e+08	4.3693e+07	0	
...	...	...	...	
P22599	4.2849e+06	2.4446e+07	2.9112e+07	
Q64523	5.2353e+07	8.9203e+07	3.3176e+07	
F8WIT2	1.3592e+06	4.1288e+06	1.5675e+06	
Q3THW5	0	5.1085e+07	2.1225e+07	
P00920	0	9.2801e+06	1.2551e+07	

index	4DPI_cortex_3_2	4DPI_cortex_2_3	4DPI_cortex_2_2	4DPI_cortex_2_1	\
Protein IDs					
A8DUK4	6.8564e+09	8.5305e+09	8.4491e+09	3.4106e+09	
P63260	1.8186e+09	1.5937e+09	2.1845e+09	1.5055e+09	
Q91VB8	1.9416e+09	2.0437e+09	2.8438e+09	2.0836e+09	
P15864	2.659e+08	3.1331e+08	2.7457e+08	3.3855e+08	
Q8BFZ3	487500	1.944e+06	511750	404450	
...	...	...	...	...	
P22599	3.2685e+07	2.5324e+07	3.483e+07	1.8534e+07	
Q64523	3.0386e+07	1.1554e+08	4.493e+07	1.6208e+07	
F8WIT2	4.3106e+06	1.0905e+07	1.2193e+07	2.284e+06	
Q3THW5	1.4959e+07	3.1109e+07	2.8054e+07	2.7033e+06	
P00920	9.1094e+06	4.3395e+07	3.22e+07	1.4914e+07	

index	...	10DPI_cortex_3_2	10DPI_SAC_4_3	10DPI_SAC_4_2	10DPI_SAC_4_1	\
Protein IDs	...					
A8DUK4	...	8.122e+09	5.3066e+08	9.5197e+08	6.2291e+08	
P63260	...	1.3877e+09	4.8895e+09	4.6028e+09	4.8811e+09	
Q91VB8	...	5.3114e+09	3.0269e+08	4.4618e+08	3.8388e+08	
P15864	...	3.7537e+08	8.0293e+08	1.3161e+09	1.2945e+09	
Q8BFZ3	...	2.1978e+06	4.975e+08	802690	4.6505e+06	
...	...	...	...	...	...	
P22599	...	3.1196e+07	2.005e+07	1.6024e+07	1.3344e+07	
Q64523	...	5.4759e+07	6.4041e+07	1.1079e+08	5.712e+07	
F8WIT2	...	1.9329e+06	2.8045e+07	2.4982e+07	2.3535e+07	
Q3THW5	...	1.7866e+07	2.0146e+07	2.7599e+07	2.7402e+07	
P00920	...	1.0463e+07	2.3787e+06	2.5503e+06	0	

index	10DPI_SAC_3_3	10DPI_SAC_3_2	10DPI_SAC_3_1	10DPI_SAC_2_3	\
Protein IDs					
A8DUK4	8.5632e+08	8.6326e+08	9.2512e+08	1.0979e+09	

P63260	6.279e+09	6.4949e+09	7.117e+09	4.7165e+09
Q91VB8	1.0176e+09	8.2537e+08	6.0648e+08	1.3203e+09
P15864	2.0244e+09	1.426e+09	1.6338e+09	3.6126e+09
Q8BFZ3	2.016e+07	1.8383e+07	1.1963e+06	0
...	...	...	...	...
P22599	1.1475e+07	9.8682e+06	1.4829e+07	1.6949e+06
Q64523	6.8067e+07	6.8305e+07	6.2162e+07	7.2188e+07
F8WIT2	2.866e+07	4.5345e+07	4.2387e+07	1.0955e+07
Q3THW5	4.4057e+07	4.8592e+07	3.2142e+07	3.4779e+07
P00920	3.199e+06	5.1937e+06	5.0533e+06	1.5782e+06

index	10DPI_SAC_2_2	10DPI_SAC_2_1
Protein IDs		
A8DUK4	1.2606e+09	1.8661e+09
P63260	4.8012e+09	4.7378e+09
Q91VB8	2.0021e+09	8.6326e+08
P15864	1.3788e+09	2.4812e+09
Q8BFZ3	5.581e+06	4.5486e+08
...	...	...
P22599	1.2092e+07	6.6191e+06
Q64523	1.1346e+08	5.8851e+07
F8WIT2	1.5971e+07	4.3004e+06
Q3THW5	5.4393e+07	1.1939e+07
P00920	2.9578e+06	989490

[175 rows x 39 columns]

```
[80]: df2 = df.transpose()
df2 = df2.astype('int64')
```

```
[81]: df3 = StandardScaler().fit_transform(df2)
df3 = pd.DataFrame(df3, columns=df2.columns, index=df2.index)
df3
```

```
[81]: Protein IDs      A8DUK4      P63260      Q91VB8      P15864      Q8BFZ3  \
index
4DPI_interface_2_4 -0.540090  0.021508 -0.022581 -0.236924  1.759115
4DPI_interface_2_2  0.019054 -0.337894  0.113558 -0.642223 -0.428211
4DPI_interface_2_1 -0.138874 -0.540108 -0.215675 -0.984486 -0.420023
4DPI_interface_1_2  0.608371 -1.105209 -0.325806 -0.810748  0.163989
4DPI_interface_1_1  0.174370 -1.183182  0.094234 -0.679225 -0.370694
4DPI_cortex_3_3     0.812077 -1.342470  0.739405 -1.104028 -0.428877
4DPI_cortex_3_2     1.171869 -1.082481  0.131356 -1.092429 -0.428228
4DPI_cortex_2_3     1.732487 -1.189867  0.205528 -1.029867 -0.426288
4DPI_cortex_2_2     1.705228 -0.907770  0.786771 -1.080988 -0.428195
4DPI_cortex_2_1     0.017949 -1.231981  0.234514 -0.996560 -0.428338
4DPI_cortex_1_2     3.162245 -0.851188  2.487419 -0.892101  1.517026
4DPI_cortex_1_1     2.284197 -1.246783  2.586436 -1.133441  1.497051
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4DPI_SAC_3_3	-0.653982	0.011098	-0.271394	1.321765	-0.413909
4DPI_SAC_3_2	-0.555428	0.006037	-0.709916	0.867429	-0.428290
4DPI_SAC_2_3	-0.512329	-0.250133	-0.485555	-0.434520	-0.428877
4DPI_SAC_2_2	-0.480516	-0.370315	-0.212042	-0.525994	4.685493
4DPI_SAC_2_1	-0.371781	0.039127	-0.427148	0.498867	-0.426050
4DPI_SAC_1_3	0.162783	0.504720	0.867335	-0.745891	1.806121
4DPI_SAC_1_2	-0.281163	0.401154	2.496209	-0.308182	-0.422143
4DPI_SAC_1_1	-0.043099	0.343665	0.096850	-0.764444	-0.425772
10DPI_interface_5_3	-0.777652	1.419721	-0.753751	0.824938	-0.425504
10DPI_interface_4_3	-0.866100	1.606369	-1.090698	1.305138	-0.426965
10DPI_interface_4_2	-0.873578	2.688393	-1.119343	0.698258	-0.427497
10DPI_interface_4_1	-0.800096	1.695420	-0.998278	1.009681	-0.422012
10DPI_interface_3_2	-0.632316	0.429326	-0.690294	0.874819	-0.427889
10DPI_interface_3_1	-0.719082	1.182173	-0.586279	0.609317	-0.424988
10DPI_cortex_4_4	0.249718	-0.833378	-0.002022	-0.349340	-0.417555
10DPI_cortex_4_3	1.169458	-1.045285	0.320527	-0.289708	-0.428877
10DPI_cortex_3_3	0.492771	-1.152337	0.040476	-1.045689	0.826432
10DPI_cortex_3_2	1.595690	-1.288228	2.579389	-0.947973	-0.425950
10DPI_SAC_4_3	-0.946477	0.383822	-1.059250	-0.383768	0.233606
10DPI_SAC_4_2	-0.805390	0.246927	-0.955010	0.293407	-0.427808
10DPI_SAC_4_1	-0.915585	0.379811	-1.000268	0.264904	-0.422684
10DPI_SAC_3_3	-0.837421	1.047284	-0.539895	1.228074	-0.402031
10DPI_SAC_3_2	-0.835097	1.150373	-0.679543	0.438430	-0.404398
10DPI_SAC_3_1	-0.814382	1.447415	-0.838558	0.712641	-0.427284
10DPI_SAC_2_3	-0.756522	0.301217	-0.319995	3.323850	-0.428877
10DPI_SAC_2_2	-0.702037	0.341660	0.175307	0.376145	-0.421445
10DPI_SAC_2_1	-0.499269	0.311387	-0.652017	1.830863	0.176826

Protein IDs	P31725	P20152	P10107	Q64475	Q03265	...	\
index						...	
4DPI_interface_2_4	-0.206117	-0.124103	-0.339158	0.487461	-0.225207	...	
4DPI_interface_2_2	-0.643921	-0.213255	-0.527045	-0.433295	0.089789	...	
4DPI_interface_2_1	-0.478166	-0.400444	-0.238710	-0.155440	-0.369529	...	
4DPI_interface_1_2	-0.708444	-0.569122	-1.012065	0.125845	-0.241213	...	
4DPI_interface_1_1	-0.682565	-0.517408	-0.894347	-1.043352	0.168928	...	
4DPI_cortex_3_3	-0.716880	-1.326299	-1.073669	-1.645795	1.809980	...	
4DPI_cortex_3_2	-0.716880	-1.250061	-1.073243	-1.052557	1.360204	...	
4DPI_cortex_2_3	-0.716880	-1.103115	-1.059976	-1.384593	2.563874	...	
4DPI_cortex_2_2	-0.706149	-1.122011	-0.999513	-1.012349	1.560213	...	
4DPI_cortex_2_1	-0.716197	-1.377694	-1.067414	-1.569062	2.644484	...	
4DPI_cortex_1_2	-0.715155	-1.028298	-1.053593	-0.699643	0.906860	...	
4DPI_cortex_1_1	-0.716079	-1.165106	-1.051863	-1.281794	0.987024	...	
4DPI_SAC_3_3	-0.209229	0.137459	0.461878	1.972168	-0.525400	...	
4DPI_SAC_3_2	-0.467254	0.003198	0.190899	0.522564	-0.485897	...	
4DPI_SAC_2_3	-0.670474	-0.619947	-0.568406	0.362906	-0.125960	...	
4DPI_SAC_2_2	-0.716077	-0.592846	-1.023940	1.084758	-0.046063	...	
4DPI_SAC_2_1	-0.611808	0.228388	0.345181	-0.117659	-0.187086	...	

4DPI_SAC_1_3	0.169334	-0.339045	-0.138673	-0.801061	-0.217939	...
4DPI_SAC_1_2	0.438196	-0.292751	0.361676	-0.780309	-0.500477	...
4DPI_SAC_1_1	0.151373	-0.312506	0.113881	-0.266356	-0.382281	...
10DPI_interface_5_3	3.100625	1.692848	1.770771	1.013422	-0.901580	...
10DPI_interface_4_3	0.123680	1.805398	1.579930	1.870624	-0.827514	...
10DPI_interface_4_2	-0.433562	1.568746	1.171000	1.245463	-0.759250	...
10DPI_interface_4_1	0.143423	1.820800	1.449857	0.584194	-0.808740	...
10DPI_interface_3_2	2.058175	1.296107	1.581119	1.029865	-0.899163	...
10DPI_interface_3_1	2.024052	1.189421	1.322902	0.444367	-0.726168	...
10DPI_cortex_4_4	-0.708326	-0.885863	-0.975620	-0.368067	0.463414	...
10DPI_cortex_4_3	-0.714705	-0.986684	-0.967459	-0.910387	0.495828	...
10DPI_cortex_3_3	-0.716880	-1.388084	-1.074705	-1.033478	0.805517	...
10DPI_cortex_3_2	-0.713643	-1.307077	-1.065171	-1.234055	1.836598	...
10DPI_SAC_4_3	-0.566626	0.631378	-0.248722	-0.012392	-0.887629	...
10DPI_SAC_4_2	-0.557917	1.028207	-0.059768	-0.264724	-0.928505	...
10DPI_SAC_4_1	-0.457542	0.442797	0.157376	0.156136	-0.804905	...
10DPI_SAC_3_3	1.704192	1.220225	1.609022	0.738915	-0.772893	...
10DPI_SAC_3_2	1.818186	0.917108	1.475871	0.665529	-0.810746	...
10DPI_SAC_3_1	1.932369	0.795435	2.036087	1.164043	-0.729155	...
10DPI_SAC_2_3	0.561078	0.472652	0.058406	2.246173	-0.941055	...
10DPI_SAC_2_2	0.716264	1.252213	0.695106	0.680466	-0.734416	...
10DPI_SAC_2_1	0.326527	0.419339	0.132100	-0.328529	-0.853945	...

Protein IDs index	Q3UZ39	Q8K2B3	P54071	Q5EBP8	Q3UHL6	\
4DPI_interface_2_4	0.009676	-0.563738	-0.572222	1.176545	0.479523	
4DPI_interface_2_2	-0.480505	0.312642	0.108868	-0.112737	1.020549	
4DPI_interface_2_1	-0.124023	-0.195956	-0.288919	-0.287406	3.358514	
4DPI_interface_1_2	-1.111607	-0.847701	-1.313029	2.526452	-1.398293	
4DPI_interface_1_1	-0.794028	-0.183291	-0.396191	-0.633064	-0.763125	
4DPI_cortex_3_3	-1.111607	2.389069	2.780540	-1.209991	-1.200441	
4DPI_cortex_3_2	-1.075394	2.642528	1.103917	-1.254383	-1.116769	
4DPI_cortex_2_3	-1.111607	0.289798	0.747838	-1.480287	-0.601693	
4DPI_cortex_2_2	-0.985139	0.927562	2.473702	-1.063374	-0.635564	
4DPI_cortex_2_1	-1.111607	2.781043	0.885157	-1.124328	-1.457122	
4DPI_cortex_1_2	-1.042013	0.402159	0.387543	-0.610224	-1.050960	
4DPI_cortex_1_1	-1.111607	0.372535	1.688444	-1.307895	-0.087518	
4DPI_SAC_3_3	0.859422	-0.118965	-0.215371	1.173940	-0.019713	
4DPI_SAC_3_2	0.670442	-0.205360	0.332292	0.644875	0.366192	
4DPI_SAC_2_3	-0.910960	-0.847701	0.709180	-0.148376	0.388922	
4DPI_SAC_2_2	-0.731997	-0.804805	-0.295466	0.099180	0.356018	
4DPI_SAC_2_1	0.775021	-0.144981	-0.312418	-0.697214	1.143217	
4DPI_SAC_1_3	-0.264832	-0.231101	-0.275555	-0.319366	0.441080	
4DPI_SAC_1_2	-0.049937	-0.791810	-0.465452	-0.401526	0.663750	
4DPI_SAC_1_1	-0.011629	-0.563415	-0.294121	-0.452111	0.864976	
10DPI_interface_5_3	1.366443	-0.589826	-0.831299	2.509054	-0.034459	
10DPI_interface_4_3	2.075350	-0.607916	-0.539646	0.123322	0.360525	



10DPI_interface_4_2	1.317175	-0.534558	-0.955586	-0.056711	0.403089
10DPI_interface_4_1	2.387756	-0.684575	-0.765294	-0.331015	0.798845
10DPI_interface_3_2	1.252235	-0.518486	-0.647940	0.174290	0.137984
10DPI_interface_3_1	0.535851	-0.799975	-0.656345	0.932209	0.511461
10DPI_cortex_4_4	-1.111607	0.561984	0.415706	-0.251920	-1.130942
10DPI_cortex_4_3	-1.111607	0.144019	0.267265	-0.745192	-1.740476
10DPI_cortex_3_3	-0.989523	2.103080	2.198077	-0.986310	-1.499564
10DPI_cortex_3_2	-1.028619	1.714496	1.500717	-1.132283	-1.404205
10DPI_SAC_4_3	0.139248	-0.309391	-0.591542	-0.022605	1.212825
10DPI_SAC_4_2	1.237076	-0.215797	-0.877033	-0.136190	0.888029
10DPI_SAC_4_1	0.181858	-0.695181	-0.943190	0.053271	0.811080
10DPI_SAC_3_3	1.177360	-0.528697	-0.646102	0.962406	0.188661
10DPI_SAC_3_2	0.354553	-0.653498	-0.791834	1.968034	0.611656
10DPI_SAC_3_1	1.073498	-0.685842	-0.365068	1.402105	0.516033
10DPI_SAC_2_3	0.803599	-0.695569	-0.758504	1.101512	-0.351013
10DPI_SAC_2_2	-0.537363	-0.807827	-0.990091	-0.199726	-0.260928
10DPI_SAC_2_1	0.590650	-0.814955	-0.811028	0.117038	-0.770144

Protein IDs	P22599	Q64523	F8WIT2	Q3THW5	P00920
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4DPI_interface_2_4	1.060029	-0.806552	0.333282	0.265110	0.825307
4DPI_interface_2_2	0.448626	0.748592	-0.255785	-0.275693	-0.570818
4DPI_interface_2_1	1.273251	-1.326470	1.400135	-0.002805	0.761545
4DPI_interface_1_2	-1.298396	-0.549035	-1.284947	-2.020523	-0.954043
4DPI_interface_1_1	0.569832	1.132906	-1.074569	0.933596	-0.258719
4DPI_cortex_3_3	1.002206	-1.424328	-1.269124	-0.793134	-0.013643
4DPI_cortex_3_2	1.333298	-1.551672	-1.060759	-1.155481	-0.271509
4DPI_cortex_2_3	0.651192	2.335002	-0.559851	-0.221566	2.297385
4DPI_cortex_2_2	1.532065	-0.887842	-0.462015	-0.398230	1.458584
4DPI_cortex_2_1	0.021997	-2.198797	-1.214699	-1.864198	0.163408
4DPI_cortex_1_2	1.854168	-1.613975	-0.514503	-0.225499	0.553250
4DPI_cortex_1_1	2.578901	-0.316210	-0.125894	1.446816	3.128243
4DPI_SAC_3_3	-0.660388	-1.006833	-0.533417	1.745495	-0.678786
4DPI_SAC_3_2	-1.468260	1.232316	-0.343139	-0.897570	-0.639428
4DPI_SAC_2_3	0.612272	-0.370115	-0.942338	0.166398	-0.074633
4DPI_SAC_2_2	-0.560032	0.430142	-0.712401	1.940489	-0.120488
4DPI_SAC_2_1	0.853479	1.009259	-0.392664	1.248641	-0.060097
4DPI_SAC_1_3	1.264170	0.645348	0.203771	0.802443	1.905520
4DPI_SAC_1_2	-0.210963	-0.418359	-0.223654	1.123733	1.281908
4DPI_SAC_1_1	0.141256	0.209230	0.059371	1.953616	2.253478
10DPI_interface_5_3	-0.716358	0.069289	2.108762	-0.058551	0.101444
10DPI_interface_4_3	-0.542148	-0.052531	1.100246	-0.278758	-0.738001
10DPI_interface_4_2	-0.794289	0.402985	0.692647	0.261583	-0.605321
10DPI_interface_4_1	-1.136797	-0.539039	0.663326	0.134073	-0.315896
10DPI_interface_3_2	-0.844300	0.683414	1.966034	0.753521	-0.302776
10DPI_interface_3_1	-0.834747	-0.058054	1.545065	-0.882188	-0.520317
10DPI_cortex_4_4	-0.326331	0.770136	-0.645693	-0.487515	-0.636610

10DPI_cortex_4_3	-1.315724	-0.145369	-0.752614	0.355841	-0.561474
10DPI_cortex_3_3	-0.246361	-0.325248	-0.998138	-1.806480	-0.441809
10DPI_cortex_3_2	1.195320	-0.439218	-1.241369	-0.987376	-0.170089
10DPI_SAC_4_3	0.162477	-0.015561	0.742096	-0.855530	-0.775816
10DPI_SAC_4_2	-0.210593	2.118198	0.509432	-0.424541	-0.762959
10DPI_SAC_4_1	-0.458935	-0.331455	0.399519	-0.435933	-0.954043
10DPI_SAC_3_3	-0.632125	0.168198	0.788812	0.527184	-0.714354
10DPI_SAC_3_2	-0.781019	0.179061	2.056198	0.789432	-0.564898
10DPI_SAC_3_1	-0.321327	-0.101324	1.831509	-0.161831	-0.575418
10DPI_SAC_2_3	-1.538398	0.356292	-0.556053	-0.009339	-0.835794
10DPI_SAC_2_2	-0.574951	2.240065	-0.175040	1.124890	-0.732426
10DPI_SAC_2_1	-1.082097	-0.252447	-1.061534	-1.330120	-0.879904

[39 rows x 175 columns]

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[83]: kmeans = pd.read_csv('outs/' + date + '_combined_df_segmn_pca_kmeans.csv',
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kmeans
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[83]: G3UZW7;A0A023T778;Q9CQL1;P61327 \

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4DPI_cortex_2_2	0.0
4DPI_cortex_2_1	0.0
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4DPI_SAC_2_2	0.0
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4DPI_SAC_1_2	779590.0
4DPI_SAC_1_1	0.0
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10DPI_interface_4_1	0.0
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10DPI_SAC_4_3	0.0
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070589;A0A067XG53;F6Y9I5 \

index

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4DPI_SAC_1_1	0.0
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10DPI_interface_4_3	0.0
10DPI_interface_4_2	0.0
10DPI_interface_4_1	0.0
10DPI_interface_3_2	0.0
10DPI_interface_3_1	0.0
10DPI_cortex_4_4	0.0
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B2;AOA140T8N1;AOA0G2JDE5;P01631 \

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4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
4DPI_SAC_1_2	0.0
4DPI_SAC_1_1	0.0
10DPI_interface_5_3	0.0
10DPI_interface_4_3	0.0
10DPI_interface_4_2	0.0
10DPI_interface_4_1	0.0
10DPI_interface_3_2	0.0
10DPI_interface_3_1	0.0
10DPI_cortex_4_4	0.0
10DPI_cortex_4_3	0.0
10DPI_cortex_3_3	855310.0
10DPI_cortex_3_2	498950.0
10DPI_SAC_4_3	1674100.0
10DPI_SAC_4_2	431080.0
10DPI_SAC_4_1	548100.0
10DPI_SAC_3_3	0.0
10DPI_SAC_3_2	0.0
10DPI_SAC_3_1	0.0

10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	0.0
10DPI_SAC_2_1	0.0

A0A075B5M7;A0A0G2JDV4;A0A0B4J1J2;A0A0B4J1J1 \

index

4DPI_interface_2_4	0.0
4DPI_interface_2_2	0.0
4DPI_interface_2_1	0.0
4DPI_interface_1_2	0.0
4DPI_interface_1_1	0.0
4DPI_cortex_3_3	0.0
4DPI_cortex_3_2	0.0
4DPI_cortex_2_3	0.0
4DPI_cortex_2_2	0.0
4DPI_cortex_2_1	0.0
4DPI_cortex_1_2	0.0
4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
4DPI_SAC_1_2	0.0
4DPI_SAC_1_1	0.0
10DPI_interface_5_3	81477.0
10DPI_interface_4_3	207140.0
10DPI_interface_4_2	0.0
10DPI_interface_4_1	0.0
10DPI_interface_3_2	0.0
10DPI_interface_3_1	0.0
10DPI_cortex_4_4	0.0
10DPI_cortex_4_3	0.0
10DPI_cortex_3_3	0.0
10DPI_cortex_3_2	0.0
10DPI_SAC_4_3	202930.0
10DPI_SAC_4_2	290690.0
10DPI_SAC_4_1	0.0
10DPI_SAC_3_3	0.0
10DPI_SAC_3_2	0.0
10DPI_SAC_3_1	0.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	0.0
10DPI_SAC_2_1	0.0

A0A075B5P3;A0A0A6YVP0;P01867 \

index	
4DPI_interface_2_4	0.0
4DPI_interface_2_2	0.0
4DPI_interface_2_1	0.0
4DPI_interface_1_2	0.0
4DPI_interface_1_1	0.0
4DPI_cortex_3_3	0.0
4DPI_cortex_3_2	0.0
4DPI_cortex_2_3	0.0
4DPI_cortex_2_2	0.0
4DPI_cortex_2_1	0.0
4DPI_cortex_1_2	0.0
4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
4DPI_SAC_1_2	0.0
4DPI_SAC_1_1	0.0
10DPI_interface_5_3	0.0
10DPI_interface_4_3	1183700.0
10DPI_interface_4_2	2295600.0
10DPI_interface_4_1	1475700.0
10DPI_interface_3_2	869990.0
10DPI_interface_3_1	0.0
10DPI_cortex_4_4	0.0
10DPI_cortex_4_3	0.0
10DPI_cortex_3_3	0.0
10DPI_cortex_3_2	0.0
10DPI_SAC_4_3	1207600.0
10DPI_SAC_4_2	1526100.0
10DPI_SAC_4_1	1165600.0
10DPI_SAC_3_3	0.0
10DPI_SAC_3_2	0.0
10DPI_SAC_3_1	724260.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	0.0
10DPI_SAC_2_1	0.0

AOA075B5P5;AOA1Y7VJN6;P03987 \

index	
4DPI_interface_2_4	0.0
4DPI_interface_2_2	0.0
4DPI_interface_2_1	0.0
4DPI_interface_1_2	0.0

4DPI_interface_1_1	0.0
4DPI_cortex_3_3	0.0
4DPI_cortex_3_2	0.0
4DPI_cortex_2_3	0.0
4DPI_cortex_2_2	0.0
4DPI_cortex_2_1	0.0
4DPI_cortex_1_2	0.0
4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0
4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
4DPI_SAC_1_2	0.0
4DPI_SAC_1_1	0.0
10DPI_interface_5_3	1109600.0
10DPI_interface_4_3	12624000.0
10DPI_interface_4_2	14386000.0
10DPI_interface_4_1	19824000.0
10DPI_interface_3_2	3005900.0
10DPI_interface_3_1	447980.0
10DPI_cortex_4_4	21601000.0
10DPI_cortex_4_3	526710.0
10DPI_cortex_3_3	1102700.0
10DPI_cortex_3_2	654170.0
10DPI_SAC_4_3	29513000.0
10DPI_SAC_4_2	13752000.0
10DPI_SAC_4_1	17690000.0
10DPI_SAC_3_3	848950.0
10DPI_SAC_3_2	997060.0
10DPI_SAC_3_1	295760.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	1941400.0
10DPI_SAC_2_1	0.0

AOA075B6A0;AOA075B5P6;P01872 \

index

4DPI_interface_2_4	11850000.0
4DPI_interface_2_2	13198000.0
4DPI_interface_2_1	14041000.0
4DPI_interface_1_2	8327300.0
4DPI_interface_1_1	9088300.0
4DPI_cortex_3_3	28479000.0
4DPI_cortex_3_2	9475600.0
4DPI_cortex_2_3	12548000.0
4DPI_cortex_2_2	17613000.0

4DPI_cortex_2_1	18376000.0
4DPI_cortex_1_2	13749000.0
4DPI_cortex_1_1	19070000.0
4DPI_SAC_3_3	4038600.0
4DPI_SAC_3_2	1051500.0
4DPI_SAC_2_3	10500000.0
4DPI_SAC_2_2	4068800.0
4DPI_SAC_2_1	6212200.0
4DPI_SAC_1_3	7820100.0
4DPI_SAC_1_2	3859700.0
4DPI_SAC_1_1	4501600.0
10DPI_interface_5_3	7111900.0
10DPI_interface_4_3	15394000.0
10DPI_interface_4_2	20355000.0
10DPI_interface_4_1	24453000.0
10DPI_interface_3_2	6101300.0
10DPI_interface_3_1	9250300.0
10DPI_cortex_4_4	49938000.0
10DPI_cortex_4_3	16417000.0
10DPI_cortex_3_3	16583000.0
10DPI_cortex_3_2	14249000.0
10DPI_SAC_4_3	36836000.0
10DPI_SAC_4_2	21015000.0
10DPI_SAC_4_1	20078000.0
10DPI_SAC_3_3	5357600.0
10DPI_SAC_3_2	2988700.0
10DPI_SAC_3_1	6025900.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	6863300.0
10DPI_SAC_2_1	2568900.0

AOA075B5R2;AOA075B5R3;AOA0A6YX91;P01786 \

index

4DPI_interface_2_4	489990.0
4DPI_interface_2_2	0.0
4DPI_interface_2_1	1280300.0
4DPI_interface_1_2	0.0
4DPI_interface_1_1	0.0
4DPI_cortex_3_3	0.0
4DPI_cortex_3_2	0.0
4DPI_cortex_2_3	537080.0
4DPI_cortex_2_2	1492000.0
4DPI_cortex_2_1	247720.0
4DPI_cortex_1_2	0.0
4DPI_cortex_1_1	0.0
4DPI_SAC_3_3	0.0
4DPI_SAC_3_2	0.0



4DPI_SAC_2_3	0.0
4DPI_SAC_2_2	0.0
4DPI_SAC_2_1	0.0
4DPI_SAC_1_3	0.0
4DPI_SAC_1_2	0.0
4DPI_SAC_1_1	0.0
10DPI_interface_5_3	923480.0
10DPI_interface_4_3	2742000.0
10DPI_interface_4_2	2199900.0
10DPI_interface_4_1	2193900.0
10DPI_interface_3_2	875140.0
10DPI_interface_3_1	1436900.0
10DPI_cortex_4_4	1717800.0
10DPI_cortex_4_3	1395700.0
10DPI_cortex_3_3	880010.0
10DPI_cortex_3_2	708140.0
10DPI_SAC_4_3	1401500.0
10DPI_SAC_4_2	0.0
10DPI_SAC_4_1	1453700.0
10DPI_SAC_3_3	536050.0
10DPI_SAC_3_2	2143600.0
10DPI_SAC_3_1	944670.0
10DPI_SAC_2_3	0.0
10DPI_SAC_2_2	533920.0
10DPI_SAC_2_1	0.0

AOA075B5S5;AOA0A6YXL5 AOA075B5T3;J3QK03;AOA0A6YWS9 ... \

index			
4DPI_interface_2_4	0.0	0.0	...
4DPI_interface_2_2	0.0	0.0	...
4DPI_interface_2_1	0.0	0.0	...
4DPI_interface_1_2	0.0	0.0	...
4DPI_interface_1_1	0.0	0.0	...
4DPI_cortex_3_3	0.0	0.0	...
4DPI_cortex_3_2	0.0	0.0	...
4DPI_cortex_2_3	0.0	0.0	...
4DPI_cortex_2_2	0.0	0.0	...
4DPI_cortex_2_1	0.0	0.0	...
4DPI_cortex_1_2	0.0	0.0	...
4DPI_cortex_1_1	0.0	0.0	...
4DPI_SAC_3_3	0.0	0.0	...
4DPI_SAC_3_2	0.0	0.0	...
4DPI_SAC_2_3	0.0	0.0	...
4DPI_SAC_2_2	0.0	0.0	...
4DPI_SAC_2_1	0.0	0.0	...
4DPI_SAC_1_3	0.0	0.0	...
4DPI_SAC_1_2	0.0	0.0	...

4DPI_SAC_1_1	0.0	0.0	...
10DPI_interface_5_3	582000.0	624130.0	...
10DPI_interface_4_3	10474000.0	1520300.0	...
10DPI_interface_4_2	4267600.0	3319600.0	...
10DPI_interface_4_1	1899800.0	1112100.0	...
10DPI_interface_3_2	521140.0	826650.0	...
10DPI_interface_3_1	554440.0	0.0	...
10DPI_cortex_4_4	3546300.0	2818100.0	...
10DPI_cortex_4_3	0.0	2965700.0	...
10DPI_cortex_3_3	0.0	992900.0	...
10DPI_cortex_3_2	739180.0	0.0	...
10DPI_SAC_4_3	2447600.0	5502600.0	...
10DPI_SAC_4_2	3972000.0	2063300.0	...
10DPI_SAC_4_1	8814000.0	2556300.0	...
10DPI_SAC_3_3	0.0	1867700.0	...
10DPI_SAC_3_2	0.0	766080.0	...
10DPI_SAC_3_1	0.0	0.0	...
10DPI_SAC_2_3	0.0	0.0	...
10DPI_SAC_2_2	0.0	0.0	...
10DPI_SAC_2_1	0.0	0.0	...

	36	37	38 \
index			
4DPI_interface_2_4	-9.667796e+06	3.778624e+07	2.949520e-07
4DPI_interface_2_2	4.484831e+07	3.021682e+07	2.949520e-07
4DPI_interface_2_1	-3.927658e+07	-2.821519e+07	2.949520e-07
4DPI_interface_1_2	1.602163e+07	1.157634e+07	2.949520e-07
4DPI_interface_1_1	-2.771801e+07	-1.684883e+07	2.949520e-07
4DPI_cortex_3_3	-2.290089e+07	1.351457e+07	2.949520e-07
4DPI_cortex_3_2	1.978959e+07	-2.313470e+07	2.949520e-07
4DPI_cortex_2_3	6.588185e+06	-3.016993e+07	2.949520e-07
4DPI_cortex_2_2	-2.862014e+07	5.637718e+07	2.949520e-07
4DPI_cortex_2_1	3.341529e+05	2.571811e+05	2.949520e-07
4DPI_cortex_1_2	-4.056160e+07	-1.501557e+07	2.949520e-07
4DPI_cortex_1_1	7.268118e+07	-1.174222e+07	2.949520e-07
4DPI_SAC_3_3	-5.645220e+06	3.757987e+07	2.949520e-07
4DPI_SAC_3_2	3.419064e+07	-4.386457e+07	2.949520e-07
4DPI_SAC_2_3	2.145412e+07	1.723141e+06	2.949520e-07
4DPI_SAC_2_2	-2.183335e+06	-1.551289e+07	2.949520e-07
4DPI_SAC_2_1	-1.893998e+07	-5.739152e+06	2.949520e-07
4DPI_SAC_1_3	-2.059662e+07	1.852929e+07	2.949520e-07
4DPI_SAC_1_2	3.062973e+06	-1.035837e+07	2.949520e-07
4DPI_SAC_1_1	-1.611010e+06	3.801341e+06	2.949520e-07
10DPI_interface_5_3	1.634717e+07	-9.292562e+04	2.949520e-07
10DPI_interface_4_3	-2.843180e+07	-7.313157e+06	2.949520e-07
10DPI_interface_4_2	7.446369e+06	-2.292092e+06	2.949520e-07
10DPI_interface_4_1	2.750186e+07	2.790485e+06	2.949520e-07

10DPI_interface_3_2	1.381678e+07	2.345572e+07	2.949520e-07
10DPI_interface_3_1	-3.987217e+07	-2.743997e+07	2.949520e-07
10DPI_cortex_4_4	-7.014139e+06	-1.328333e+07	2.949520e-07
10DPI_cortex_4_3	5.219814e+06	2.369728e+07	2.949520e-07
10DPI_cortex_3_3	6.556227e+06	-7.602979e+06	2.949520e-07
10DPI_cortex_3_2	-4.925120e+06	4.877634e+05	2.949520e-07
10DPI_SAC_4_3	8.374328e+05	-5.950123e+06	2.949520e-07
10DPI_SAC_4_2	-9.686167e+06	2.496185e+07	2.949520e-07
10DPI_SAC_4_1	3.640798e+06	-1.856121e+07	2.949520e-07
10DPI_SAC_3_3	-2.675078e+06	-1.110516e+07	2.949520e-07
10DPI_SAC_3_2	-2.930084e+06	-3.858284e+06	2.949520e-07
10DPI_SAC_3_1	2.071382e+07	1.198856e+07	2.949520e-07
10DPI_SAC_2_3	-7.880436e+06	-1.545762e+07	2.949520e-07
10DPI_SAC_2_2	-2.200700e+06	-1.017980e+07	2.949520e-07
10DPI_SAC_2_1	2.285816e+06	2.499442e+07	2.949520e-07

	Segment	K-means	PCA	TimePoint	BioRep	TechRep	\
index							
4DPI_interface_2_4		1		4DPI	2	4	
4DPI_interface_2_2		4		4DPI	2	2	
4DPI_interface_2_1		4		4DPI	2	1	
4DPI_interface_1_2		4		4DPI	1	2	
4DPI_interface_1_1		4		4DPI	1	1	
4DPI_cortex_3_3		0		4DPI	3	3	
4DPI_cortex_3_2		0		4DPI	3	2	
4DPI_cortex_2_3		0		4DPI	2	3	
4DPI_cortex_2_2		0		4DPI	2	2	
4DPI_cortex_2_1		4		4DPI	2	1	
4DPI_cortex_1_2		2		4DPI	1	2	
4DPI_cortex_1_1		2		4DPI	1	1	
4DPI_SAC_3_3		1		4DPI	3	3	
4DPI_SAC_3_2		1		4DPI	3	2	
4DPI_SAC_2_3		1		4DPI	2	3	
4DPI_SAC_2_2		1		4DPI	2	2	
4DPI_SAC_2_1		1		4DPI	2	1	
4DPI_SAC_1_3		1		4DPI	1	3	
4DPI_SAC_1_2		1		4DPI	1	2	
4DPI_SAC_1_1		1		4DPI	1	1	
10DPI_interface_5_3		3		10DPI	5	3	
10DPI_interface_4_3		3		10DPI	4	3	
10DPI_interface_4_2		3		10DPI	4	2	
10DPI_interface_4_1		3		10DPI	4	1	
10DPI_interface_3_2		1		10DPI	3	2	
10DPI_interface_3_1		3		10DPI	3	1	
10DPI_cortex_4_4		4		10DPI	4	4	
10DPI_cortex_4_3		0		10DPI	4	3	
10DPI_cortex_3_3		4		10DPI	3	3	

10DPI_cortex_3_2	0	10DPI	3	2
10DPI_SAC_4_3	1	10DPI	4	3
10DPI_SAC_4_2	1	10DPI	4	2
10DPI_SAC_4_1	1	10DPI	4	1
10DPI_SAC_3_3	3	10DPI	3	3
10DPI_SAC_3_2	3	10DPI	3	2
10DPI_SAC_3_1	3	10DPI	3	1
10DPI_SAC_2_3	1	10DPI	2	3
10DPI_SAC_2_2	1	10DPI	2	2
10DPI_SAC_2_1	1	10DPI	2	1

	Region	DPI_Region	BioRep_TechRep
index			
4DPI_interface_2_4	interface	4DPI_interface	2_4
4DPI_interface_2_2	interface	4DPI_interface	2_2
4DPI_interface_2_1	interface	4DPI_interface	2_1
4DPI_interface_1_2	interface	4DPI_interface	1_2
4DPI_interface_1_1	interface	4DPI_interface	1_1
4DPI_cortex_3_3	cortex	4DPI_cortex	3_3
4DPI_cortex_3_2	cortex	4DPI_cortex	3_2
4DPI_cortex_2_3	cortex	4DPI_cortex	2_3
4DPI_cortex_2_2	cortex	4DPI_cortex	2_2
4DPI_cortex_2_1	cortex	4DPI_cortex	2_1
4DPI_cortex_1_2	cortex	4DPI_cortex	1_2
4DPI_cortex_1_1	cortex	4DPI_cortex	1_1
4DPI_SAC_3_3	SAC	4DPI_SAC	3_3
4DPI_SAC_3_2	SAC	4DPI_SAC	3_2
4DPI_SAC_2_3	SAC	4DPI_SAC	2_3
4DPI_SAC_2_2	SAC	4DPI_SAC	2_2
4DPI_SAC_2_1	SAC	4DPI_SAC	2_1
4DPI_SAC_1_3	SAC	4DPI_SAC	1_3
4DPI_SAC_1_2	SAC	4DPI_SAC	1_2
4DPI_SAC_1_1	SAC	4DPI_SAC	1_1
10DPI_interface_5_3	interface	10DPI_interface	5_3
10DPI_interface_4_3	interface	10DPI_interface	4_3
10DPI_interface_4_2	interface	10DPI_interface	4_2
10DPI_interface_4_1	interface	10DPI_interface	4_1
10DPI_interface_3_2	interface	10DPI_interface	3_2
10DPI_interface_3_1	interface	10DPI_interface	3_1
10DPI_cortex_4_4	cortex	10DPI_cortex	4_4
10DPI_cortex_4_3	cortex	10DPI_cortex	4_3
10DPI_cortex_3_3	cortex	10DPI_cortex	3_3
10DPI_cortex_3_2	cortex	10DPI_cortex	3_2
10DPI_SAC_4_3	SAC	10DPI_SAC	4_3
10DPI_SAC_4_2	SAC	10DPI_SAC	4_2
10DPI_SAC_4_1	SAC	10DPI_SAC	4_1
10DPI_SAC_3_3	SAC	10DPI_SAC	3_3

10DPI_SAC_3_2	SAC	10DPI_SAC	3_2
10DPI_SAC_3_1	SAC	10DPI_SAC	3_1
10DPI_SAC_2_3	SAC	10DPI_SAC	2_3
10DPI_SAC_2_2	SAC	10DPI_SAC	2_2
10DPI_SAC_2_1	SAC	10DPI_SAC	2_1

[39 rows x 3659 columns]

```
[84]: df3['k'] = kmeans['Segment K-means PCA']
df3
```

```
[84]: Protein IDs      A8DUK4      P63260      Q91VB8      P15864      Q8BFZ3  \
index
4DPI_interface_2_4 -0.540090  0.021508 -0.022581 -0.236924  1.759115
4DPI_interface_2_2  0.019054 -0.337894  0.113558 -0.642223 -0.428211
4DPI_interface_2_1 -0.138874 -0.540108 -0.215675 -0.984486 -0.420023
4DPI_interface_1_2  0.608371 -1.105209 -0.325806 -0.810748  0.163989
4DPI_interface_1_1  0.174370 -1.183182  0.094234 -0.679225 -0.370694
4DPI_cortex_3_3     0.812077 -1.342470  0.739405 -1.104028 -0.428877
4DPI_cortex_3_2     1.171869 -1.082481  0.131356 -1.092429 -0.428228
4DPI_cortex_2_3     1.732487 -1.189867  0.205528 -1.029867 -0.426288
4DPI_cortex_2_2     1.705228 -0.907770  0.786771 -1.080988 -0.428195
4DPI_cortex_2_1     0.017949 -1.231981  0.234514 -0.996560 -0.428338
4DPI_cortex_1_2     3.162245 -0.851188  2.487419 -0.892101  1.517026
4DPI_cortex_1_1     2.284197 -1.246783  2.586436 -1.133441  1.497051
4DPI_SAC_3_3        -0.653982  0.011098 -0.271394  1.321765 -0.413909
4DPI_SAC_3_2        -0.555428  0.006037 -0.709916  0.867429 -0.428290
4DPI_SAC_2_3        -0.512329 -0.250133 -0.485555 -0.434520 -0.428877
4DPI_SAC_2_2        -0.480516 -0.370315 -0.212042 -0.525994  4.685493
4DPI_SAC_2_1        -0.371781  0.039127 -0.427148  0.498867 -0.426050
4DPI_SAC_1_3         0.162783  0.504720  0.867335 -0.745891  1.806121
4DPI_SAC_1_2        -0.281163  0.401154  2.496209 -0.308182 -0.422143
4DPI_SAC_1_1        -0.043099  0.343665  0.096850 -0.764444 -0.425772
10DPI_interface_5_3 -0.777652  1.419721 -0.753751  0.824938 -0.425504
10DPI_interface_4_3 -0.866100  1.606369 -1.090698  1.305138 -0.426965
10DPI_interface_4_2 -0.873578  2.688393 -1.119343  0.698258 -0.427497
10DPI_interface_4_1 -0.800096  1.695420 -0.998278  1.009681 -0.422012
10DPI_interface_3_2 -0.632316  0.429326 -0.690294  0.874819 -0.427889
10DPI_interface_3_1 -0.719082  1.182173 -0.586279  0.609317 -0.424988
10DPI_cortex_4_4     0.249718 -0.833378 -0.002022 -0.349340 -0.417555
10DPI_cortex_4_3     1.169458 -1.045285  0.320527 -0.289708 -0.428877
10DPI_cortex_3_3     0.492771 -1.152337  0.040476 -1.045689  0.826432
10DPI_cortex_3_2     1.595690 -1.288228  2.579389 -0.947973 -0.425950
10DPI_SAC_4_3        -0.946477  0.383822 -1.059250 -0.383768  0.233606
10DPI_SAC_4_2        -0.805390  0.246927 -0.955010  0.293407 -0.427808
10DPI_SAC_4_1        -0.915585  0.379811 -1.000268  0.264904 -0.422684
10DPI_SAC_3_3        -0.837421  1.047284 -0.539895  1.228074 -0.402031
10DPI_SAC_3_2        -0.835097  1.150373 -0.679543  0.438430 -0.404398
```

10DPI_SAC_3_1	-0.814382	1.447415	-0.838558	0.712641	-0.427284
10DPI_SAC_2_3	-0.756522	0.301217	-0.319995	3.323850	-0.428877
10DPI_SAC_2_2	-0.702037	0.341660	0.175307	0.376145	-0.421445
10DPI_SAC_2_1	-0.499269	0.311387	-0.652017	1.830863	0.176826

Protein IDs	P31725	P20152	P10107	Q64475	Q03265	...	\
index						...	
4DPI_interface_2_4	-0.206117	-0.124103	-0.339158	0.487461	-0.225207	...	
4DPI_interface_2_2	-0.643921	-0.213255	-0.527045	-0.433295	0.089789	...	
4DPI_interface_2_1	-0.478166	-0.400444	-0.238710	-0.155440	-0.369529	...	
4DPI_interface_1_2	-0.708444	-0.569122	-1.012065	0.125845	-0.241213	...	
4DPI_interface_1_1	-0.682565	-0.517408	-0.894347	-1.043352	0.168928	...	
4DPI_cortex_3_3	-0.716880	-1.326299	-1.073669	-1.645795	1.809980	...	
4DPI_cortex_3_2	-0.716880	-1.250061	-1.073243	-1.052557	1.360204	...	
4DPI_cortex_2_3	-0.716880	-1.103115	-1.059976	-1.384593	2.563874	...	
4DPI_cortex_2_2	-0.706149	-1.122011	-0.999513	-1.012349	1.560213	...	
4DPI_cortex_2_1	-0.716197	-1.377694	-1.067414	-1.569062	2.644484	...	
4DPI_cortex_1_2	-0.715155	-1.028298	-1.053593	-0.699643	0.906860	...	
4DPI_cortex_1_1	-0.716079	-1.165106	-1.051863	-1.281794	0.987024	...	
4DPI_SAC_3_3	-0.209229	0.137459	0.461878	1.972168	-0.525400	...	
4DPI_SAC_3_2	-0.467254	0.003198	0.190899	0.522564	-0.485897	...	
4DPI_SAC_2_3	-0.670474	-0.619947	-0.568406	0.362906	-0.125960	...	
4DPI_SAC_2_2	-0.716077	-0.592846	-1.023940	1.084758	-0.046063	...	
4DPI_SAC_2_1	-0.611808	0.228388	0.345181	-0.117659	-0.187086	...	
4DPI_SAC_1_3	0.169334	-0.339045	-0.138673	-0.801061	-0.217939	...	
4DPI_SAC_1_2	0.438196	-0.292751	0.361676	-0.780309	-0.500477	...	
4DPI_SAC_1_1	0.151373	-0.312506	0.113881	-0.266356	-0.382281	...	
10DPI_interface_5_3	3.100625	1.692848	1.770771	1.013422	-0.901580	...	
10DPI_interface_4_3	0.123680	1.805398	1.579930	1.870624	-0.827514	...	
10DPI_interface_4_2	-0.433562	1.568746	1.171000	1.245463	-0.759250	...	
10DPI_interface_4_1	0.143423	1.820800	1.449857	0.584194	-0.808740	...	
10DPI_interface_3_2	2.058175	1.296107	1.581119	1.029865	-0.899163	...	
10DPI_interface_3_1	2.024052	1.189421	1.322902	0.444367	-0.726168	...	
10DPI_cortex_4_4	-0.708326	-0.885863	-0.975620	-0.368067	0.463414	...	
10DPI_cortex_4_3	-0.714705	-0.986684	-0.967459	-0.910387	0.495828	...	
10DPI_cortex_3_3	-0.716880	-1.388084	-1.074705	-1.033478	0.805517	...	
10DPI_cortex_3_2	-0.713643	-1.307077	-1.065171	-1.234055	1.836598	...	
10DPI_SAC_4_3	-0.566626	0.631378	-0.248722	-0.012392	-0.887629	...	
10DPI_SAC_4_2	-0.557917	1.028207	-0.059768	-0.264724	-0.928505	...	
10DPI_SAC_4_1	-0.457542	0.442797	0.157376	0.156136	-0.804905	...	
10DPI_SAC_3_3	1.704192	1.220225	1.609022	0.738915	-0.772893	...	
10DPI_SAC_3_2	1.818186	0.917108	1.475871	0.665529	-0.810746	...	
10DPI_SAC_3_1	1.932369	0.795435	2.036087	1.164043	-0.729155	...	
10DPI_SAC_2_3	0.561078	0.472652	0.058406	2.246173	-0.941055	...	
10DPI_SAC_2_2	0.716264	1.252213	0.695106	0.680466	-0.734416	...	
10DPI_SAC_2_1	0.326527	0.419339	0.132100	-0.328529	-0.853945	...	

Protein IDs	Q8K2B3	P54071	Q5EBP8	Q3UHL6	P22599	\
index						
4DPI_interface_2_4	-0.563738	-0.572222	1.176545	0.479523	1.060029	
4DPI_interface_2_2	0.312642	0.108868	-0.112737	1.020549	0.448626	
4DPI_interface_2_1	-0.195956	-0.288919	-0.287406	3.358514	1.273251	
4DPI_interface_1_2	-0.847701	-1.313029	2.526452	-1.398293	-1.298396	
4DPI_interface_1_1	-0.183291	-0.396191	-0.633064	-0.763125	0.569832	
4DPI_cortex_3_3	2.389069	2.780540	-1.209991	-1.200441	1.002206	
4DPI_cortex_3_2	2.642528	1.103917	-1.254383	-1.116769	1.333298	
4DPI_cortex_2_3	0.289798	0.747838	-1.480287	-0.601693	0.651192	
4DPI_cortex_2_2	0.927562	2.473702	-1.063374	-0.635564	1.532065	
4DPI_cortex_2_1	2.781043	0.885157	-1.124328	-1.457122	0.021997	
4DPI_cortex_1_2	0.402159	0.387543	-0.610224	-1.050960	1.854168	
4DPI_cortex_1_1	0.372535	1.688444	-1.307895	-0.087518	2.578901	
4DPI_SAC_3_3	-0.118965	-0.215371	1.173940	-0.019713	-0.660388	
4DPI_SAC_3_2	-0.205360	0.332292	0.644875	0.366192	-1.468260	
4DPI_SAC_2_3	-0.847701	0.709180	-0.148376	0.388922	0.612272	
4DPI_SAC_2_2	-0.804805	-0.295466	0.099180	0.356018	-0.560032	
4DPI_SAC_2_1	-0.144981	-0.312418	-0.697214	1.143217	0.853479	
4DPI_SAC_1_3	-0.231101	-0.275555	-0.319366	0.441080	1.264170	
4DPI_SAC_1_2	-0.791810	-0.465452	-0.401526	0.663750	-0.210963	
4DPI_SAC_1_1	-0.563415	-0.294121	-0.452111	0.864976	0.141256	
10DPI_interface_5_3	-0.589826	-0.831299	2.509054	-0.034459	-0.716358	
10DPI_interface_4_3	-0.607916	-0.539646	0.123322	0.360525	-0.542148	
10DPI_interface_4_2	-0.534558	-0.955586	-0.056711	0.403089	-0.794289	
10DPI_interface_4_1	-0.684575	-0.765294	-0.331015	0.798845	-1.136797	
10DPI_interface_3_2	-0.518486	-0.647940	0.174290	0.137984	-0.844300	
10DPI_interface_3_1	-0.799975	-0.656345	0.932209	0.511461	-0.834747	
10DPI_cortex_4_4	0.561984	0.415706	-0.251920	-1.130942	-0.326331	
10DPI_cortex_4_3	0.144019	0.267265	-0.745192	-1.740476	-1.315724	
10DPI_cortex_3_3	2.103080	2.198077	-0.986310	-1.499564	-0.246361	
10DPI_cortex_3_2	1.714496	1.500717	-1.132283	-1.404205	1.195320	
10DPI_SAC_4_3	-0.309391	-0.591542	-0.022605	1.212825	0.162477	
10DPI_SAC_4_2	-0.215797	-0.877033	-0.136190	0.888029	-0.210593	
10DPI_SAC_4_1	-0.695181	-0.943190	0.053271	0.811080	-0.458935	
10DPI_SAC_3_3	-0.528697	-0.646102	0.962406	0.188661	-0.632125	
10DPI_SAC_3_2	-0.653498	-0.791834	1.968034	0.611656	-0.781019	
10DPI_SAC_3_1	-0.685842	-0.365068	1.402105	0.516033	-0.321327	
10DPI_SAC_2_3	-0.695569	-0.758504	1.101512	-0.351013	-1.538398	
10DPI_SAC_2_2	-0.807827	-0.990091	-0.199726	-0.260928	-0.574951	
10DPI_SAC_2_1	-0.814955	-0.811028	0.117038	-0.770144	-1.082097	
Protein IDs	Q64523	F8WIT2	Q3THW5	P00920	k	
index						
4DPI_interface_2_4	-0.806552	0.333282	0.265110	0.825307	1	
4DPI_interface_2_2	0.748592	-0.255785	-0.275693	-0.570818	4	
4DPI_interface_2_1	-1.326470	1.400135	-0.002805	0.761545	4	

4DPI_interface_1_2	-0.549035	-1.284947	-2.020523	-0.954043	4
4DPI_interface_1_1	1.132906	-1.074569	0.933596	-0.258719	4
4DPI_cortex_3_3	-1.424328	-1.269124	-0.793134	-0.013643	0
4DPI_cortex_3_2	-1.551672	-1.060759	-1.155481	-0.271509	0
4DPI_cortex_2_3	2.335002	-0.559851	-0.221566	2.297385	0
4DPI_cortex_2_2	-0.887842	-0.462015	-0.398230	1.458584	0
4DPI_cortex_2_1	-2.198797	-1.214699	-1.864198	0.163408	4
4DPI_cortex_1_2	-1.613975	-0.514503	-0.225499	0.553250	2
4DPI_cortex_1_1	-0.316210	-0.125894	1.446816	3.128243	2
4DPI_SAC_3_3	-1.006833	-0.533417	1.745495	-0.678786	1
4DPI_SAC_3_2	1.232316	-0.343139	-0.897570	-0.639428	1
4DPI_SAC_2_3	-0.370115	-0.942338	0.166398	-0.074633	1
4DPI_SAC_2_2	0.430142	-0.712401	1.940489	-0.120488	1
4DPI_SAC_2_1	1.009259	-0.392664	1.248641	-0.060097	1
4DPI_SAC_1_3	0.645348	0.203771	0.802443	1.905520	1
4DPI_SAC_1_2	-0.418359	-0.223654	1.123733	1.281908	1
4DPI_SAC_1_1	0.209230	0.059371	1.953616	2.253478	1
10DPI_interface_5_3	0.069289	2.108762	-0.058551	0.101444	3
10DPI_interface_4_3	-0.052531	1.100246	-0.278758	-0.738001	3
10DPI_interface_4_2	0.402985	0.692647	0.261583	-0.605321	3
10DPI_interface_4_1	-0.539039	0.663326	0.134073	-0.315896	3
10DPI_interface_3_2	0.683414	1.966034	0.753521	-0.302776	1
10DPI_interface_3_1	-0.058054	1.545065	-0.882188	-0.520317	3
10DPI_cortex_4_4	0.770136	-0.645693	-0.487515	-0.636610	4
10DPI_cortex_4_3	-0.145369	-0.752614	0.355841	-0.561474	0
10DPI_cortex_3_3	-0.325248	-0.998138	-1.806480	-0.441809	4
10DPI_cortex_3_2	-0.439218	-1.241369	-0.987376	-0.170089	0
10DPI_SAC_4_3	-0.015561	0.742096	-0.855530	-0.775816	1
10DPI_SAC_4_2	2.118198	0.509432	-0.424541	-0.762959	1
10DPI_SAC_4_1	-0.331455	0.399519	-0.435933	-0.954043	1
10DPI_SAC_3_3	0.168198	0.788812	0.527184	-0.714354	3
10DPI_SAC_3_2	0.179061	2.056198	0.789432	-0.564898	3
10DPI_SAC_3_1	-0.101324	1.831509	-0.161831	-0.575418	3
10DPI_SAC_2_3	0.356292	-0.556053	-0.009339	-0.835794	1
10DPI_SAC_2_2	2.240065	-0.175040	1.124890	-0.732426	1
10DPI_SAC_2_1	-0.252447	-1.061534	-1.330120	-0.879904	1

[39 rows x 176 columns]

```
[85]: df3['k'] = df3['k'].astype('category')
df3.dtypes
```

```
[85]: Protein IDs
A8DUK4      float64
P63260      float64
Q91VB8      float64
P15864      float64
Q8BFZ3      float64
```



```

...
Q64523      float64
F8WIT2      float64
Q3THW5      float64
P00920      float64
k           category
Length: 176, dtype: object

```

```
[86]: df4 = df3.groupby(['k']).mean()
df4
```

```
[86]: Protein IDs      A8DUK4      P63260      Q91VB8      P15864      Q8BFZ3      P31725  \
k
0           1.364468 -1.142684  0.793830 -0.924165 -0.427736 -0.714189
1          -0.533325  0.193813 -0.198111  0.390770  0.249214 -0.002631
2           2.723221 -1.048986  2.536927 -1.012771  1.507039 -0.715617
3          -0.815426  1.529644 -0.825793  0.853310 -0.420085  1.301621
4           0.203337 -0.912013 -0.008674 -0.786896 -0.153486 -0.664928

Protein IDs      P20152      P10107      Q64475      Q03265  ...      Q3UZ39      Q8K2B3  \
k
0          -1.182541 -1.039839 -1.206623  1.604449  ... -1.070662  1.351245
1           0.226909  0.107435  0.373217 -0.546620  ...  0.250782 -0.520568
2          -1.096702 -1.052728 -0.990719  0.946942  ... -1.076810  0.387347
3           1.376248  1.551930  0.965820 -0.792006  ...  1.285998 -0.635611
4          -0.764553 -0.827130 -0.639550  0.508770  ... -0.817557  0.647400

Protein IDs      P54071      Q5EBP8      Q3UHL6      P22599      Q64523      F8WIT2  \
k
0           1.478997 -1.147585 -1.116524  0.733060 -0.352238 -0.890956
1          -0.438029  0.135221  0.396987 -0.219702  0.357684 -0.045421
2           1.037993 -0.959059 -0.569239  2.216534 -0.965092 -0.320199
3          -0.693897  0.938675  0.419476 -0.719851  0.008573  1.348321
4           0.229953 -0.124187 -0.267141  0.063231 -0.249702 -0.581957

Protein IDs      Q3THW5      P00920
k
0          -0.533324  0.456542
1           0.448206 -0.034434
2           0.610658  1.840746
3           0.041368 -0.491595
4          -0.789088 -0.276721

```

```
[5 rows x 175 columns]
```

```
[87]: df5 = df4[df4>0]
df5
```

```
[87]: Protein IDs      A8DUK4      P63260      Q91VB8      P15864      Q8BFZ3      P31725  \
k
0          1.364468          NaN      0.793830          NaN          NaN          NaN
1          NaN      0.193813          NaN      0.39077      0.249214          NaN
2          2.723221          NaN      2.536927          NaN      1.507039          NaN
3          NaN      1.529644          NaN      0.85331          NaN      1.301621
4          0.203337          NaN          NaN          NaN          NaN          NaN

Protein IDs      P20152      P10107      Q64475      Q03265      ...      Q3UZ39      Q8K2B3  \
k
0          NaN          NaN          NaN      1.604449      ...          NaN      1.351245
1          0.226909      0.107435      0.373217          NaN      ...      0.250782          NaN
2          NaN          NaN          NaN      0.946942      ...          NaN      0.387347
3          1.376248      1.551930      0.965820          NaN      ...      1.285998          NaN
4          NaN          NaN          NaN      0.508770      ...          NaN      0.647400

Protein IDs      P54071      Q5EBP8      Q3UHL6      P22599      Q64523      F8WIT2  \
k
0          1.478997          NaN          NaN      0.733060          NaN          NaN
1          NaN      0.135221      0.396987          NaN      0.357684          NaN
2          1.037993          NaN          NaN      2.216534          NaN          NaN
3          NaN      0.938675      0.419476          NaN      0.008573      1.348321
4          0.229953          NaN          NaN      0.063231          NaN          NaN

Protein IDs      Q3THW5      P00920
k
0          NaN      0.456542
1          0.448206          NaN
2          0.610658      1.840746
3          0.041368          NaN
4          NaN          NaN

[5 rows x 175 columns]
```

```
[88]: proteinIDs = []

for i in range(0,5):
    proteinList = df5.iloc[i][-df5.iloc[i].isnull()].index
    proteinIDs.append(proteinList)

proteinIDs
```

```
[88]: [Index(['A8DUK4', 'Q91VB8', 'Q03265', 'Q91Y97', 'P31786', 'P56480', 'P16125',
            'Q8VDN2', 'Q9JII6', 'Q99KIO', 'P63038', 'Q9DCW4', 'P13745', 'P14094',
            'P26443', 'A2ARV4', 'Q64433', 'Q61838', 'P99029', 'G5E8R3', 'P08249',
            'Q62468', 'P24270', 'Q64442', 'Q61646', 'O88844', 'Q9EQ20', 'Q9DBM2',
            'Q9D881', 'Q9Z2I8', 'P45952', 'P70441', 'P97450', 'P17751', 'Q8K0L3',
            'Q921I1', 'A3KGU5', 'P38647', 'P51881', 'P56391', 'P14152',
```

```

        'AOA0R4J1V0', 'Q62433', 'P10922', 'Q8BWT1', 'E9PZF0', 'P05202',
        'AOA1L1SV25', 'Q9DB20', 'F8WIP8', 'O08749', 'P48758', 'Q60932',
        'Q9DCX2', 'P16460', 'P07724', 'P17182', 'P56135', 'Q62261', 'Q91X72',
        'Q8VC30', 'Q9QXD6', 'Q8QZT1', 'P52760', 'Q9JLJ2', 'P47199', 'Q8ROY6',
        'Q8CHT0', 'Q8K2B3', 'P54071', 'P22599', 'P00920'],
        dtype='object', name=' Protein IDs'),
Index(['P63260', 'P15864', 'Q8BFZ3', 'P20152', 'P10107', 'Q64475', 'P27661',
        'P43276', 'P20065', 'P52480', 'P43274', 'Q8VDD5', 'B7FAV1', 'O35744',
        'Q61176', 'AOA0N4SV66', 'P43277', 'P26041', 'P62962', 'E9PV24',
        'P43275', 'P05064', 'P26350', 'P09528', 'Q61233', 'P27773', 'D3Z2H9',
        'P63017', 'P20029', 'P26039', 'O08692', 'P07356', 'P49710', 'P18760',
        'Q99PT1', 'O88569', 'O89053', 'P08071', 'Q99LB4', 'E9Q7Q3', 'Q61599',
        'Q91Z25', 'P40124', 'E9Q3W4', 'P40142', 'P13020', 'P11247', 'P30681',
        'P11499', 'A2AL12', 'E9Q616', 'P14733', 'Q6IRU2', 'Q8KOE8', 'Q3UER8',
        'Q8BH61', 'P19973', 'P05213', 'P17742', 'Q9DCD0', 'P26645', 'P06745',
        'P56391', 'P63101', 'Q9JKF1', 'P24527', 'Q61029', 'AOA1W2P6F6',
        'P61979', 'AOA2R8VHH0', 'O35639', 'AOA0R4J0I1', 'P08113', 'P99024',
        'P09103', 'Q00612', 'Q8BMK4', 'P29351', 'Q9ET01', 'P17182', 'Q9DBJ1',
        'Q6KCD5', 'Q3TML0', 'P62960', 'Q9ZOE6', 'Q8VEK3', 'AOA1BOGR11',
        'Q05816', 'P19324', 'Q3UZ39', 'Q5EBP8', 'Q3UHL6', 'Q64523', 'Q3THW5'],
        dtype='object', name=' Protein IDs'),
Index(['A8DUK4', 'Q91VB8', 'Q8BFZ3', 'Q03265', 'Q91Y97', 'P31786', 'P56480',
        'P16125', 'Q8VDN2', 'Q9JII6', 'P26350', 'Q99KI0', 'D3Z2H9', 'P63038',
        'Q9DCW4', 'Q99PT1', 'P13745', 'P14094', 'P26443', 'A2ARV4', 'P68033',
        'Q64433', 'Q61838', 'P99029', 'AOA1D5RLD8', 'G5E8R3', 'P08249',
        'Q62468', 'P24270', 'Q64442', 'Q61646', 'O88844', 'Q9EQ20', 'Q9DBM2',
        'Q9D881', 'Q9Z2I8', 'P45952', 'P97450', 'P05213', 'P17751', 'P17742',
        'Q8KOL3', 'Q921I1', 'A3KGU5', 'P38647', 'P51881', 'P10126', 'P14152',
        'AOA0R4J1V0', 'Q62433', 'Q8BWT1', 'E9PZF0', 'P05202', 'AOA1L1SV25',
        'P61979', 'Q9DB20', 'AOA0R4J0I1', 'F8WIP8', 'O08749', 'P48758',
        'Q60932', 'Q9DCX2', 'P16460', 'P07724', 'P56135', 'Q62261', 'Q91X72',
        'Q8VC30', 'Q9QXD6', 'Q8QZT1', 'P62960', 'P52760', 'Q9JLJ2', 'P47199',
        'Q8ROY6', 'Q8CHT0', 'Q8K2B3', 'P54071', 'P22599', 'Q3THW5', 'P00920'],
        dtype='object', name=' Protein IDs'),
Index(['P63260', 'P15864', 'P31725', 'P20152', 'P10107', 'Q64475', 'P27661',
        'P43276', 'P20065', 'P52480', 'P43274', 'Q8VDD5', 'B7FAV1', 'O35744',
        'Q61176', 'AOA0N4SV66', 'P43277', 'P26041', 'P62962', 'P43275',
        'P05064', 'P62806', 'P27005', 'P09528', 'Q61233', 'P27773', 'D3Z2H9',
        'P63017', 'P20029', 'P26039', 'O08692', 'P07356', 'P49710', 'P18760',
        'Q99PT1', 'O88569', 'AOA1W2P768', 'P68033', 'O89053', 'P08071',
        'Q99LB4', 'E9Q7Q3', 'AOA1D5RLD8', 'Q61599', 'Q91Z25', 'P40124',
        'E9Q3W4', 'P40142', 'P13020', 'P11247', 'P30681', 'P11499', 'A2AL12',
        'P14733', 'Q6IRU2', 'Q8BH61', 'P19973', 'P05213', 'P17742', 'Q9DCD0',
        'P26645', 'P06745', 'Q9CQI6', 'P63101', 'Q9JKF1', 'P10126', 'P24527',
        'Q61029', 'AOA1W2P6F6', 'P61979', 'AOA2R8VHH0', 'O35639', 'P08113',
        'P99024', 'P09103', 'Q00612', 'Q8BMK4', 'P29351', 'Q9ET01', 'P17182',
        'Q9DBJ1', 'P48036', 'Q6KCD5', 'Q3TML0', 'P62960', 'Q9ZOE6', 'Q8VEK3',

```

```

        'A0A1B0GR11', 'Q05816', 'P19324', 'Q3UZ39', 'Q5EBP8', 'Q3UHL6',
        'Q64523', 'F8WIT2', 'Q3THW5'],
        dtype='object', name=' Protein IDs'),
Index(['A8DUK4', 'Q03265', 'Q91Y97', 'Q9CQT1', 'P56480', 'E9PV24', 'P16125',
       'P62806', 'Q8VDN2', 'Q9JII6', 'Q99KI0', 'P63038', 'Q9DCW4', 'P13745',
       'P14094', 'P26443', 'A2ARV4', 'A0A1W2P768', 'Q64433', 'Q61838',
       'G5E8R3', 'P08249', 'Q62468', 'P24270', 'Q64442', 'Q61646', 'O88844',
       'Q9EQ20', 'Q9DBM2', 'Q8K0E8', 'Q3UER8', 'Q9D881', 'Q9Z2I8', 'P45952',
       'P70441', 'Q8KOL3', 'Q921I1', 'A3KGU5', 'P38647', 'P51881', 'P63101',
       'P14152', 'A0A0R4J1V0', 'Q62433', 'P10922', 'Q8BWT1', 'E9PZF0',
       'P05202', 'A0A1L1SV25', 'Q9DB20', 'A0A0R4J0I1', 'F8WIP8', 'O08749',
       'P48758', 'Q60932', 'P09103', 'Q9DCX2', 'P07724', 'P56135', 'Q62261',
       'Q91X72', 'Q8VC30', 'Q9QXD6', 'Q8QZT1', 'P52760', 'Q9JLJ2', 'P47199',
       'Q8ROY6', 'Q8CHT0', 'Q8K2B3', 'P54071', 'P22599'],
        dtype='object', name=' Protein IDs'))

```

```

[89]: test = pd.DataFrame(proteinIDs)

test.to_csv('outs/' + date + '_topProteins_175.csv')

```

**0.4.2 Run proteins from line above in PANTHER and generate gene ontology classifications, then proceed with code below to combine GO information per cluster into one dataframe**

```

[90]: path = 'outs/'

fileNames = ['/*proteinClass_175.txt']
geneOntologyTerm = ['ProteinClass']

for file, go in zip(fileNames, geneOntologyTerm):

    files = glob.glob(path + file)

    li = []
    clusterID = []

    for f in files:
        c = f.split('/')[1].split('_')[0]
        clusterID.append(c)

    for filename, c in zip(files, clusterID):
        df = pd.read_csv(filename, index_col = None, header = None, sep = '\t')
        df['cluster'] = str(c)
        li.append(df)

    finalDf = pd.concat(li, axis=0, ignore_index=True).sort_values('cluster')

```

```

finalDf.columns = ['idx', 'Category name (Accession)', '# genes', 'Percent of_
→gene hit against total # genes', 'Percent of gene hit against total #_
→Protein Class hits', 'cluster']

finalDf.to_csv('outs/' + date + '_panther_combined_gene_ontology_' +_
→str(go) + '_175.csv')

```

[91]: *#save enrichment scores*

```

geneOntologyTerm = ['ProteinClass']

for go in geneOntologyTerm:
    df = pd.read_csv('outs/panther_combined_gene_ontology_' + str(go) + '_175.
→csv', index_col = 0)
    df = df[['Category name (Accession)', '# genes', 'cluster']]
    df = df.pivot(index = 'Category name (Accession)', columns = 'cluster')
    df['# genes'].to_csv('outs/' + date + '_numberGenes_' + str(go) + '_175.
→csv')

```

## 0.5 Extract LFQ intensity data for top proteins based on highest k-means centroid values

[93]:

```

summed = allProt.abs().sum(axis=1)
sorted_mylist = sorted(((v, i) for i, v in enumerate(summed)), reverse=True)[0:
→35]
final = [lis[1] for lis in sorted_mylist]
finalID = allProt.iloc[final].index
finalID

#get LFQ intensity data for top proteins based on highest k-means centroid_
→values
df = lfq.loc[finalID]
df.to_csv('outs/' + date + '_topProteinLFQ.csv')

```

[ ]: