```
In [103]: from scgenome import tantalus
          import pandas as pd
          from IPython.display import display
          from scgenome import utils, cncluster, simulation, cnplot
          import scipy.cluster.hierarchy as sch
          import matplotlib.pyplot as plt
          import numpy as np
          import seaborn as sns
          import time
          import sklearn.metrics as skm
          def get mixture labels(cn data, obs name="bhc cluster id", exp name="ori
          gin id int"):
              sub cn_data = cn_data[["cell_id", obs_name, exp_name]].drop_duplicat
          es()
              return sub_cn_data
In [75]: all cn data fp = "/Users/massoudmaher/data/sc 1935 1936 1937 cn data qc.
          csv"
          all cn_data = pd.read csv(all cn_data fp)
          all cn data = all cn data.iloc[:,1:]
          ###
          all cn data = all cn data[all cn data["chr"]=="2"]
In [88]: hmmcopy tickets = ['SC-1935', 'SC-1936', 'SC-1937']
          sample ids = [["SA922"], ['SA921'], ['SA1090']]
          # spike in params
          total ncells = 100
          proportions = [0.3, 0.3, 0.4]
          # bhc params
          n \text{ states} = 12
          alpha = 0.00001
          prob cn change = 0.7
          bhc incon = 2 # inconsistent score used for making clusters from bhc
          bhc depth = 2
          # naive clusering params
          naive method = "complete"
          naive metric = "cityblock"
          naive incon = 1.1
          naive depth = 2
          # Params for testing threshold values
          params = simulation.expand_grid({"transform":["log","none"], "criterion"
          : ["inconsistent"], "threshold": np.arange(0.025, 2, step=0.05)})
          params = pd.concat([params, simulation.expand grid({"transform":["log",
          "none"], "criterion": ["distance"], "threshold": np.arange(3, 20, step=1
          ) } ) ] )
```

```
In [77]: subsample = utils.get cn data submixture(all_cn data, total_ncells, hmmc
         opy tickets, sample ids, proportions=proportions)
         mixed_cn_data = subsample["mixed_cn_data"]
         mixed cn data["origin id int"] = mixed cn data["origin id"].factorize()[
         01
         cell_counts = subsample["cell_counts"]
         /Users/massoudmaher/Documents/Code/scgenome/scgenome/utils.py:169: Sett
         ingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row indexer,col indexer] = value instead
         See the caveats in the documentation: http://pandas.pydata.org/pandas-d
         ocs/stable/indexing.html#indexing-view-versus-copy
           jira_cn_data[origin_field_name] = hmmcopy_tickets[i]
In [89]: start = time.time()
         bhc linkage, bhc root, bhc cell ids, matrix data, measurement, variances
         = (
             cncluster.bayesian_cluster(mixed_cn_data, n_states=n_states, alpha=a
         lpha, prob cn change=prob cn change)
         print(f"{time.time()-start}s for BHC on {total_ncells} cells")
```

20.293147802352905s for BHC on 100 cells

```
In [90]: bhc_linkage, bhc_plot_data = simulation.get_plot_data(bhc_linkage)
    lbhc_plot_data = bhc_plot_data.copy()
    lbhc_plot_data[:,2] = np.log(lbhc_plot_data[:,2]) # Log because the high
    est link is way higher

naive_linkage = sch.linkage(measurement, method=naive_method, metric=nai
    ve_metric)
    ##
    naive_linkage[:,2] = naive_linkage[:,2] + 1
    lnaive_linkage = naive_linkage.copy()
    lnaive_linkage[:,2] = np.log(lnaive_linkage[:,2])
```

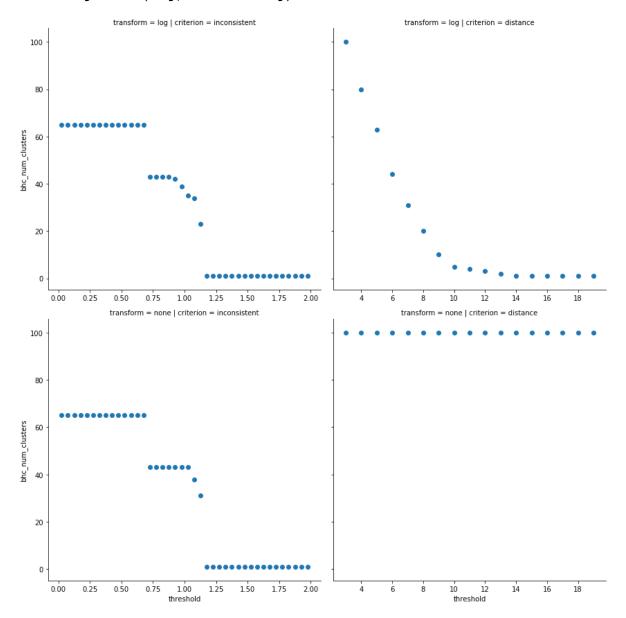
```
In [91]: def apply_fn(row):
             if row["transform"] == "log":
                 df = lbhc_plot_data
             else:
                 df = bhc_plot_data
             return sch.fcluster(df, row["threshold"], criterion=row["criterion"
         ])
         params["bhc_fcluster"] = params.apply(apply_fn, axis=1)
         params["bhc_num_clusters"] = params["bhc_fcluster"].apply(lambda x: len(
         set(x)))
         def apply_fn(row):
             if row["transform"] == "log":
                 df = lnaive linkage
             else:
                 df = naive_linkage
             return sch.fcluster(df, row["threshold"], criterion=row["criterion"
         ])
         params["naive_fcluster"] = params.apply(apply_fn, axis=1)
         params["naive_num_clusters"] = params["naive_fcluster"].apply(lambda x:
         len(set(x)))
         params.head()
```

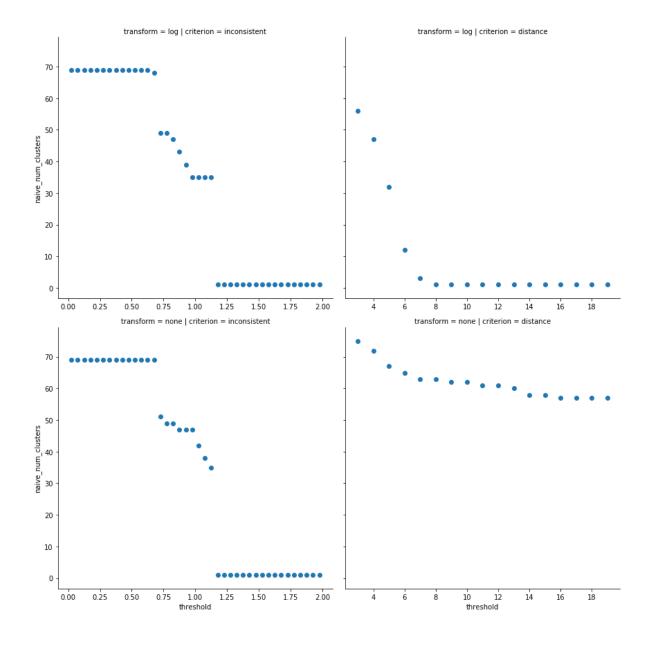
Out[91]:

	transform	criterion	threshold	bhc_fcluster	bhc_num_clusters	naive_fcluster	naive_num_cl
0	log	inconsistent	0.025	[1, 11, 19, 21, 22, 6, 13, 20, 15, 26, 17, 18,	65	[26, 21, 21, 19, 20, 27, 23, 35, 23, 69, 24, 2	
1	log	inconsistent	0.075	[1, 11, 19, 21, 22, 6, 13, 20, 15, 26, 17, 18,	65	[26, 21, 21, 19, 20, 27, 23, 35, 23, 69, 24, 2	
2	log	inconsistent	0.125	[1, 11, 19, 21, 22, 6, 13, 20, 15, 26, 17, 18,	65	[26, 21, 21, 19, 20, 27, 23, 35, 23, 69, 24, 2	
3	log	inconsistent	0.175	[1, 11, 19, 21, 22, 6, 13, 20, 15, 26, 17, 18,	65	[26, 21, 21, 19, 20, 27, 23, 35, 23, 69, 24, 2	
4	log	inconsistent	0.225	[1, 11, 19, 21, 22, 6, 13, 20, 15, 26, 17, 18,	65	[26, 21, 21, 19, 20, 27, 23, 35, 23, 69, 24, 2	

/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/seaborn/axisgrid.py:230: UserWarning: The `size` paramter has been renamed to `height`; please update your code.

warnings.warn(msg, UserWarning)





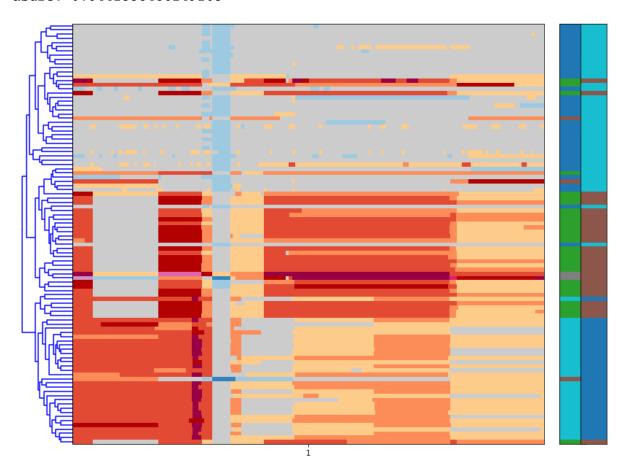
Rightmost bar represents where sample originally came frome

```
In [106]: cmixed_cn_data = mixed_cn_data.copy()
    clustering = sch.fcluster(lbhc_plot_data, 10, criterion="distance")
    cmixed_cn_data = cncluster.prune_cluster(clustering, bhc_cell_ids, mixed
    _cn_data)

fig = plt.figure(figsize=(10, 8))
    bimatrix_data = cnplot.plot_clustered_cell_cn_matrix_figure(
        fig, cmixed_cn_data, "state", cluster_field_name="bhc_cluster_id",
        linkage=lbhc_plot_data, origin_field_name="origin_id_int")

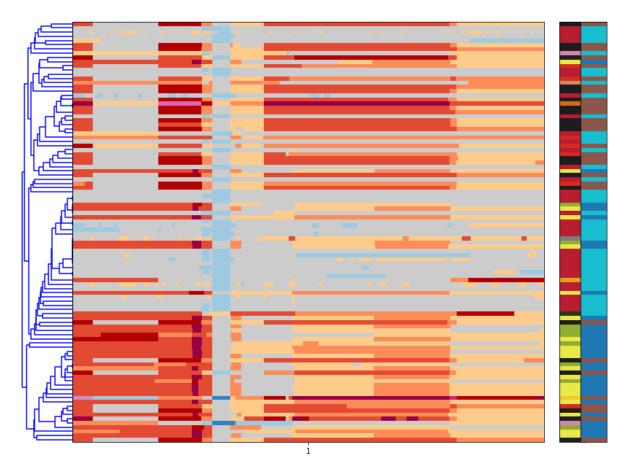
clabels = get_mixture_labels(cmixed_cn_data)
    scores = skm.homogeneity_completeness_v_measure(clabels["origin_id_int")
        l, clabels["bhc_cluster_id"])
    print(f"homogeneity: {scores[0]}, completeness: {scores[1]}, v-measure:
        {scores[2]}")
```

homogeneity: 0.9149835290813692, completeness: 0.7836400952210695, v-me asure: 0.8442338486249263



```
In [108]:
          naive_clusters = sch.fcluster(lnaive_linkage, 6, criterion="distance")
          assert len(set(naive clusters)) > 1
          nmixed on data = cncluster.prune cluster(naive clusters, bhc cell ids, m
          ixed_cn_data,
                                                   cluster_field_name="naive_clust
          er_id")
          fig = plt.figure(figsize=(10, 8))
          bimatrix data = cnplot.plot clustered_cell_cn_matrix figure(
              fig, nmixed_cn_data, "state", cluster_field_name="naive_cluster_id",
              linkage=lnaive_linkage, origin_field_name="origin_id_int")
          clabels = get mixture labels(nmixed cn data, obs name="naive cluster id"
          )
          scores = skm.homogeneity_completeness_v_measure(clabels["origin_id_int"
          ], clabels["naive_cluster_id"])
          print(f"homogeneity: {scores[0]}, completeness: {scores[1]}, v-measure:
          {scores[2]}")
```

homogeneity: 0.9609045028420629, completeness: 0.5880878934840345, v-me asure: 0.7296308313146126



```
umap_params = utils.expand_grid({"n_neighbors": np.arange(3,18,step=1)})
In [110]:
          def apply fn(row):
              return cncluster.umap hdbscan_cluster(matrix_data["state"], n_neighb
          ors=row["n_neighbors"])
          umap params["umap_clusters"] = umap_params.apply(apply_fn, axis=1)
          umap params["umap num clusters"] = umap params["umap clusters"].apply(la
          mbda x: len(set(x["cluster id"])))
          sns.scatterplot(data=umap params, x="n neighbors", y="umap num clusters"
          )
          umap df = cncluster.umap hdbscan_cluster(matrix_data["state"], n_neighbo
          rs=15)
          umixed cn data = mixed cn data.merge(umap df, how="inner")
          fig = plt.figure(figsize=(4, 4))
          cncluster.plot umap clusters(plt.gca(), umap df)
          fig = plt.figure(figsize=(4, 4))
          sns.barplot(x='cluster_id', y='count', data=umap_df.groupby('cluster_id'
          ).size().rename('count').reset_index())
          fig = plt.figure(figsize=(10, 8))
          bimatrix data = cnplot.plot clustered cell_cn matrix figure(
              fig, umixed cn data, "state", cluster field name="cluster id",
              linkage=None, origin field name="origin id int")
          clabels = get mixture labels(umixed cn data, obs name="cluster id")
          scores = skm.homogeneity completeness v measure(clabels["origin id int"
          ], clabels["cluster id"])
          print(f"homogeneity: {scores[0]}, completeness: {scores[1]}, v-measure:
          {scores[2]}")
          #def umap_hdbscan_cluster(
          #
                   cn,
                   n components=2,
          #
                   n neighbors=15,
                   min dist=0.1,
          #):
```

/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/umap/spectral.py:229: UserWarning: Embedding a total of 5 separate connected components using meta-embedding (experimental)

n components

/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/sklearn/manifold/spectral_embedding_.py:237: UserWarning: Graph is not fully connected, spectral embedding may not work as expected.

warnings.warn("Graph is not fully connected, spectral embedding" /Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/umap/spectral.py:229: UserWarning: Embedding a total of 5 separate connected components using meta-embedding (experimental)

n_components

/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/sklearn/manifold/spectral_embedding_.py:237: UserWarning: Graph is not fully connected, spectral embedding may not work as expected.

warnings.warn("Graph is not fully connected, spectral embedding" /Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/umap/spectral.py:229: UserWarning: Embedding a total of 3 separate connected components using meta-embedding (experimental)

n_components

/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/umap/spectral.py:229: UserWarning: Embedding a total of 3 separate connected components using meta-embedding (experimental)

n_components

/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/umap/spectral.py:229: UserWarning: Embedding a total of 2 separate connected components using meta-embedding (experimental)

n components

/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/umap/spectral.py:229: UserWarning: Embedding a total of 2 separate connected components using meta-embedding (experimental)

n components

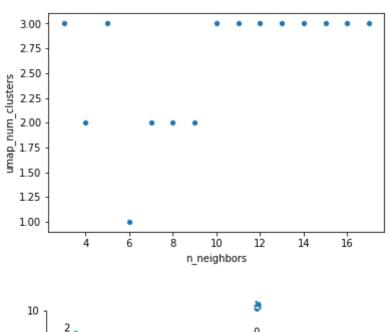
/Users/massoudmaher/Documents/Code/scgenome/scg/lib/python3.7/site-pack ages/numpy/core/fromnumeric.py:56: FutureWarning: Series.nonzero() is deprecated and will be removed in a future version.Use Series.to_numpy ().nonzero() instead

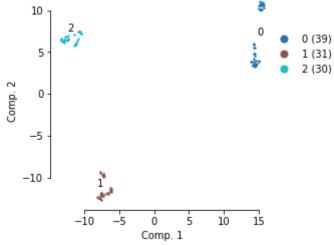
return getattr(obj, method)(*args, **kwds)

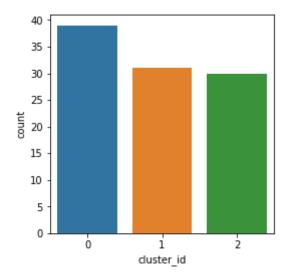
/Users/massoudmaher/Documents/Code/scgenome/scgenome/cnplot.py:50: Clus terWarning: scipy.cluster: The symmetric non-negative hollow observation matrix looks suspiciously like an uncondensed distance matrix

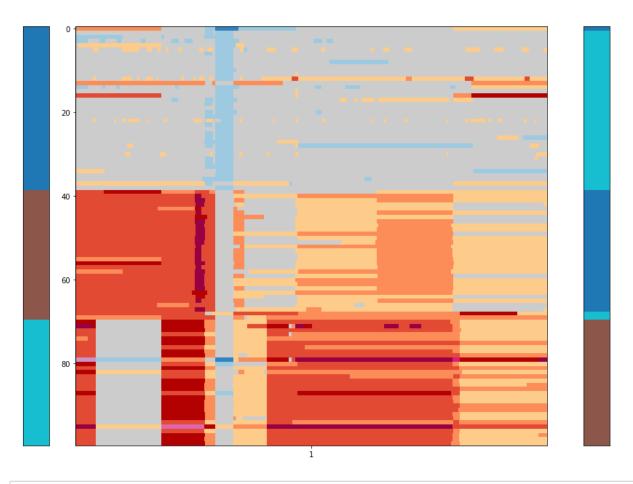
Y = sch.linkage(D, method='complete')

homogeneity: 0.8891876176786584, completeness: 0.8870809768771691, v-me asure: 0.888133048048073









In []: