

Measure 1: Find the correlation matrix between a pair of window

Figure 1 displays three correlation matrices (heatmaps) for the first, second, and last windows of the dataset. The indices are listed on the left and right of each heatmap. The first window shows high correlation between many indices, while the second and last windows show lower correlation, indicating a shift in the data distribution over time.

Step 2: Calculate Difference Score between a pair of matrices

Heatmap showing the correlation matrix of 20 protein domain families. The color scale ranges from blue (low correlation) to red (high correlation). The diagonal is red, indicating perfect self-correlation. The matrix is symmetric. The families are: M1US0007, M1USMMT, M1MUSAN, M1BRIM, M1AID, NCIDGR, M1MUEURN, M1ASOLI, M1MUNIN, M1MULPHN, M1WDSG, M1ELG, M1SLUAAJN, and M1MUAAN.

[illegible]

Measure 2: Find the largest 5 eigenvalues of each correlation matrix

corr matrix @ last window

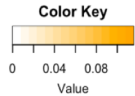
MCUAJN.Index
LSLUINDN.Index
MLCUCHNN.Index
MTASJQU.Index
MLCUAAJN.Index
MOOIEF.O.Index
MLCUBRZN.Index
MTLAIM.Index
MLCUGERN.Index
MIMUEURN.Index
MTWDLC.Index
MIMUSMT.Index
MIMUUSAN.Index

30.70
21.51
7.14
4.54
3.40

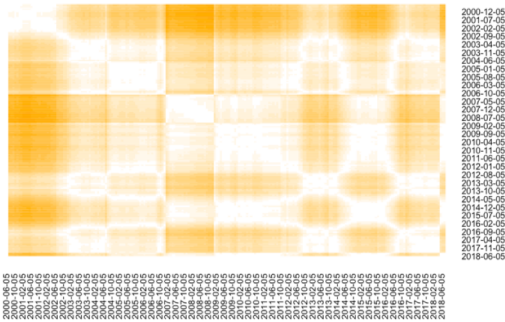
$$\sqrt{\sum |x_i - y_i|^2}$$
$$\frac{\sum x_i y_i}{\sqrt{\sum x_i^2} \sqrt{\sum y_i^2}}$$

Two Measures Heat Map Comparison

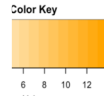
Correlation Matrix Differences



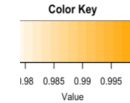
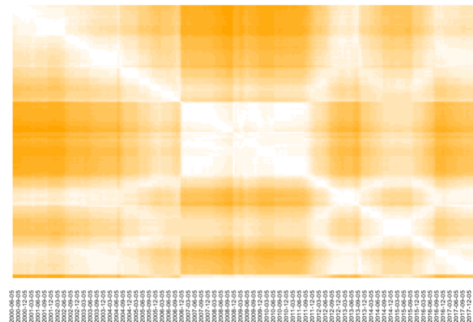
Corr Matrices Differences



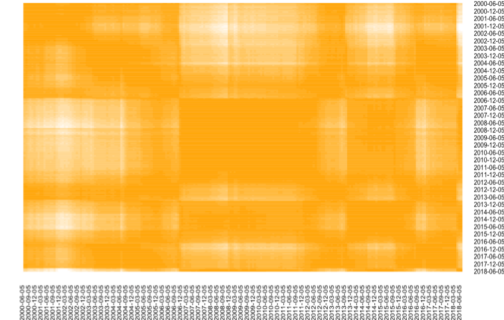
5 Largest Eigenvalue Differences



5 Largest Eigenvalues Change Using Euclidean Distance



5 Largest Eigenvalues Change Using Cosine Distance



Similar differences calculation equation

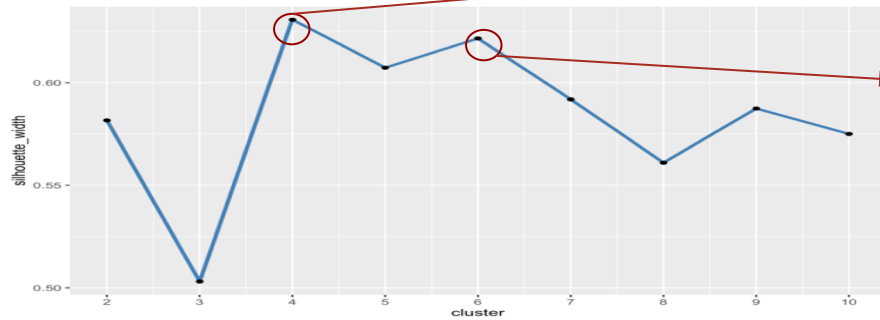
- Their scales are different
 - 5 Largest Eigenvalues Change Using Cosine Distance - **tiny**
 - Correlation matrix differences- **small**
 - 5 Largest Eigenvalues Change Using Euclidean Distance - **big**
- These three graphs look different, though the “Correlation Matrix Differences” graph and the “5 Largest Eigenvalues Change Using Euclidean Distance” graph look similar

PAM Method Clustering

PAM method

- PAM stands for “partition around medoids”. The algorithm is intended to find a sequence of objects called medoids that are centrally located in clusters.
- Input: Dissimilarity matrix
 - correlation matrix differences
 - 5 Largest Eigenvalues Cosine Distance
 - 5 Largest Eigenvalues Euclidean Distance

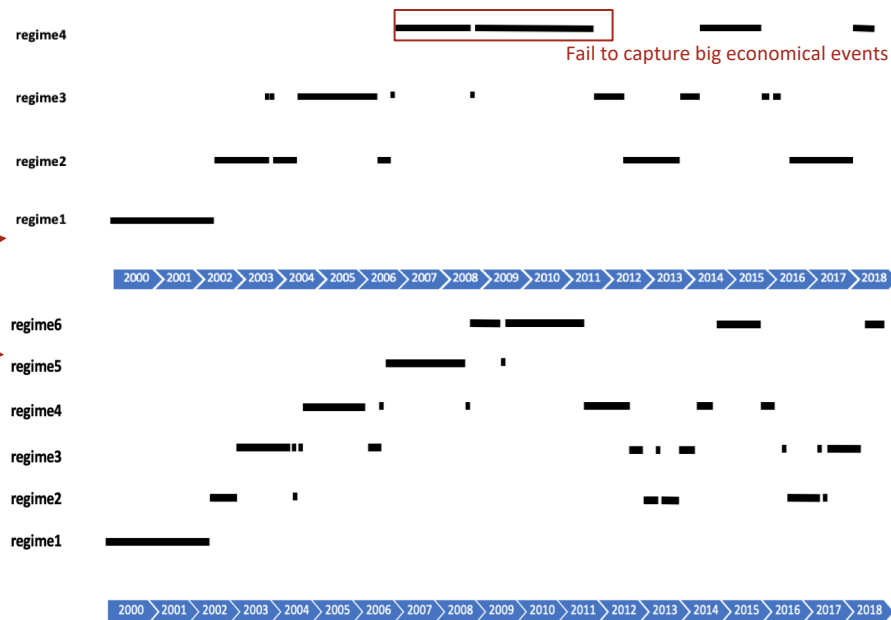
Note: use 2 yrs data first because it's preferred



Eg: Silhouette graph of Corr Matrix Diff

Silhouette analysis

- reflects how well points fit in their respective clusters.
- number of optimal clusters are 4 and 6



Optimization: A New Clustering Method Construction

Use windows as observations, and assets & factors as features

1. Within each window, find the sum of each asset&factor returns

Window1:	
Row1	(factor1, factor2, ..., factor49, asset1, asset2, ... , asset 51)
Row2	(factor1, factor2, ..., factor49, asset1, asset2, ... , asset 51)
...	
Row515	(factor1, factor2, ..., factor49, asset1, asset2, ... , asset 51)

colSum	sum factor1, ... sum factor49, sum asset1, ... sum asset51

2. Cluster on observations (windows)

Window1 (sum factor1,..., sum factor49, sum asset1, ... , sum asset 51)

Window2 (sum factor1,..., sum factor49, sum asset1, ... , sum asset 51)

...

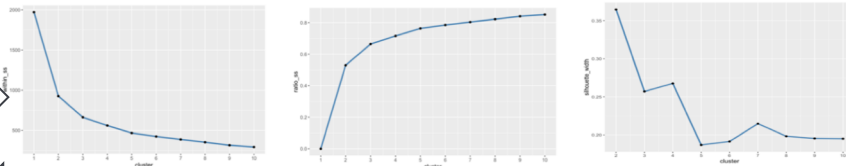
Last Window (sum factor1,...,sum factor49,sum asset1, ... ,sum asset 51)

K-means: determining optimal regime numbers by graph

Total within sum of squares Plot

Ratio Plot

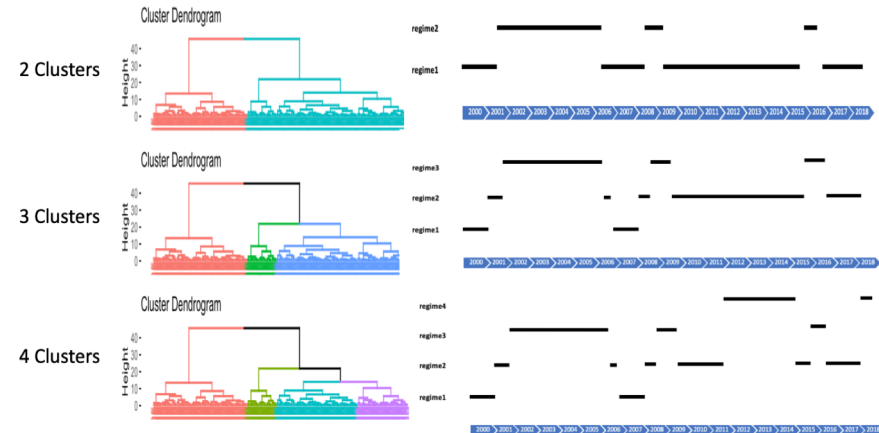
Silhouette Plot



Optimal: 3 Clusters



Hierarchical Clustering



Optimization & Regime Selection

→ 20 total market regime combinations:

4 time period (0.5 year, 1 year, 1.5 year, 2year) * 3 difference measures (corr matrix, 5 Largest Eigenvalues Cosine & Euclidean)
+ 8 (new method - k-means & hierarchical clustering for 0.5 year, 1 year, 1.5 year, 2year)

Which one is the best?

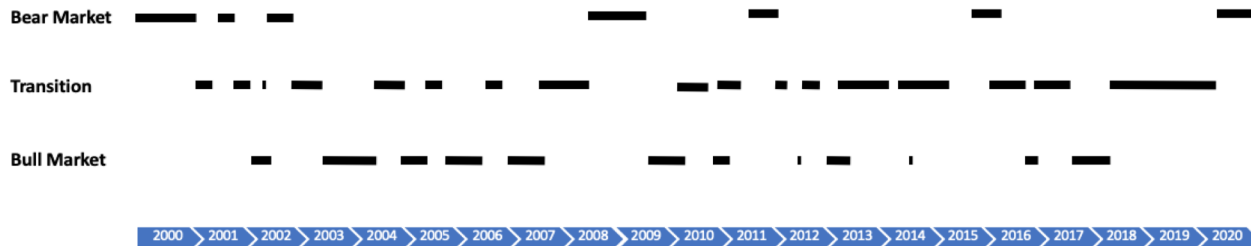
Criteria 1: Define Market Regime by Economic Cycle

- Regime should capture bull market (2002-2007 & 2009-2020) and major bear market (2000-2002, 2007-2009, 2020)

Criteria 2: Define Market Regime by volatility

- Step1: find the volatility of each assets under the same regime
- Step2: Within one regime, average all assets' volatilities (each regime has one averaged volatility value)
- Bull markets:** slow and steady, low volatility **Bear markets:** assets lose value and prices become volatile

Preferred - Optimization method, Half year, 3 Clusters



Why it's good?

- Economic life cycle: captured bear & bull
- volatility

Bear: 0.01741643 **high**

Bull: 0.0101847 **low**

Transition: 0.01045842