Portfolio
Diversification
Based on
Clustering
Analysis

Young Sim, Meng Lu, Kit Qian, Shanshan Ding, Jenny Gray





The objective of the project is to optimize stock portfolio diversification in order to reduce volatility and increase capital preservation by using clustering methods

Data Source

- Comprised of the daily close price of the 470 common stocks in S&P 500 in the 5-year data
- Date range: Feb 7, 2013 Feb 6, 2018
- Split into training period followed by test period
 - Training Period 1: Feb 2013 Feb 2017
 - Training Period 2: Feb 2016 Feb 2017
 - Test Period: Feb 2017 Feb 2018

Sharpe Ratio

- Measures how much a portfolio outperforms the risk-free rate of return on a risk-adjusted basis.
- Formula: excess return divided by standard deviation
- Assumed zero risk-free rates in all cases of the study for simplicity
- Log Return = log(1+R_i) is used for normalization purposes

Clustering

K-MEANS

- Grouping by distance to the nearest cluster center
- Assume same density and equal weights on all directions

AGGLOMERATIVE

- Start by N
 (number of data points) clusters
- Similar clusters are merged until the specified clusters are left
- Reflects hierarchy via dendrograms

DBSCAN

- Consider density of data points
- No need to specify number of clusters
- Identify noise points

Algorithm Selection

K-MEANS

Only
Euclidean
distance of
the features is
allowed in
Scikit-learn

AGGLOMERATIVE

Precomputed
Correlationbased
distance can
be used to
measure the
distance

DBSCAN

Clusters are
of very
different sizes
(no freedom
of cluster
numbers)





Algorithm Selection



Set to 30 to
match the
number of
stocks in DJI
as benchmark
for portfolio
evaluation

Linkage

- Single
- Complete
- Average

Portfolio Construction & Evaluation

DIVERSIFIED PORTFOLIO

selecting the stock with highest Sharpe ratio from each cluster with equal weight

1

Sharpe ratio in comparison to Dow Jones Index and portfolio of 30 randomly selected stocks as benchmarks

2

Annualized portfolio volatility in comparison to market volatility, 30 randomly selected stocks, and each cluster

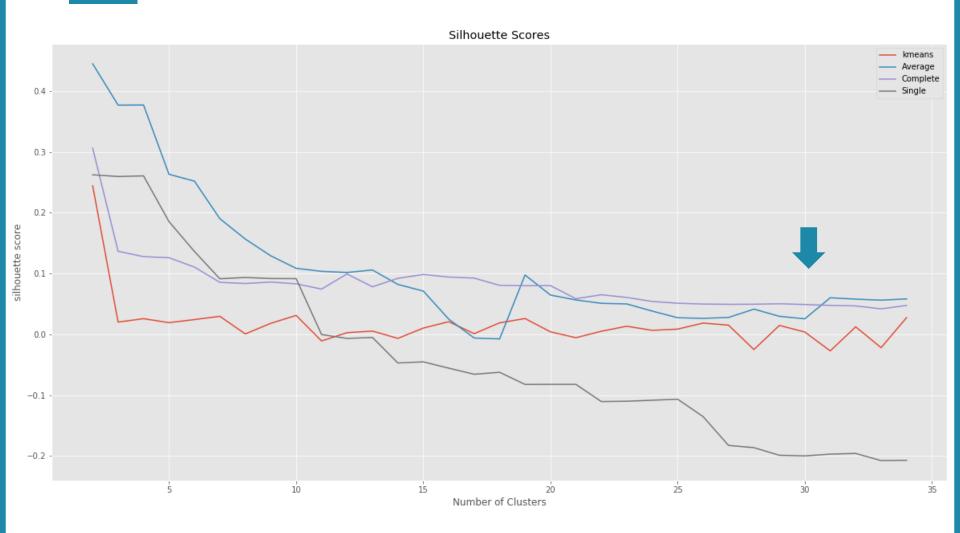
Results

Training period: Feb 7, 2013 - Feb 6, 2017

Test period: Feb 6, 2017 - Feb 6, 2018

| Benchmark /Portfolio | DJI | Random | K means | Single | Complete | Average |
|-------------------------|----------|----------|----------|---------|----------|----------|
| Sharpe Ratio | 0.150174 | 0.102065 | 0.122447 | 0.11101 | 0.085972 | 0.089122 |

Results continued

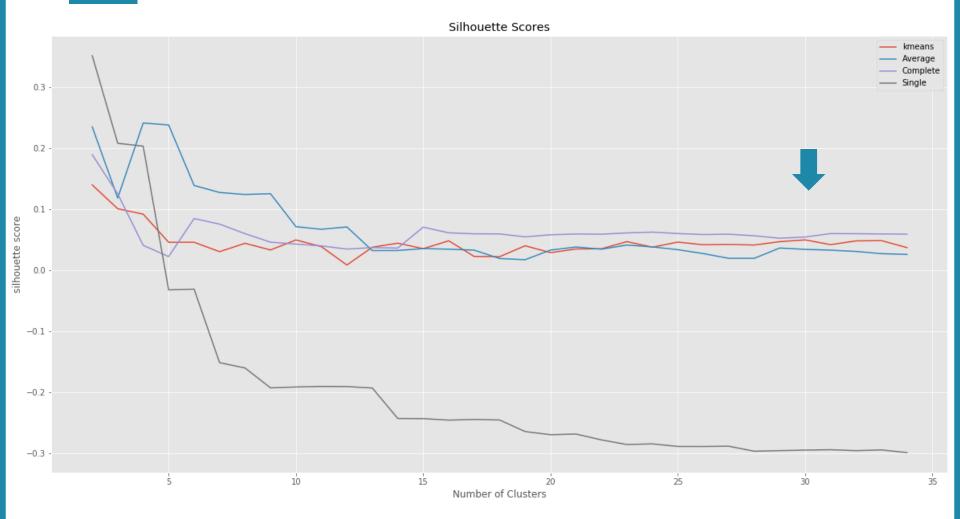


Results for Shorter Training Period

- Training period: Feb 7, 2016 Feb 6, 2017
- Test period: Feb 6, 2017 Feb 6, 2018
- Stock prices from 5 years ago may not necessarily be consistent with the prices today.

| Benchmark /Portfolio | DJI | Random | K means | Single | Complete | Average |
|-------------------------|----------|----------|----------|----------|----------|----------|
| Sharpe Ratio | 0.150174 | 0.102065 | 0.094728 | 0.075378 | 0.115475 | 0.125963 |

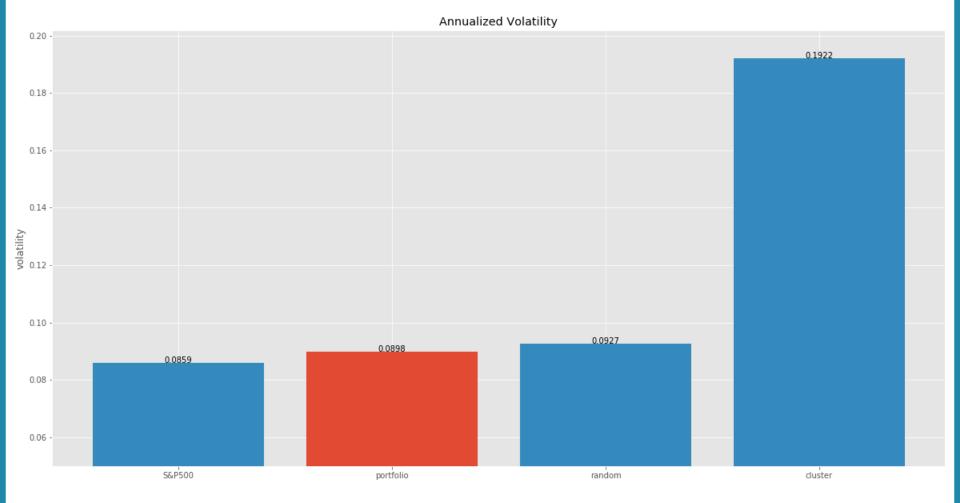
Results continued



Results continued



Only 1 out of 30 clusters had a volatility lower than the portfolio constructed.



Further Research

- 1. Other metrics can be used as alternatives of Sharpe Ratio: Sortino Ratio, Value-at-Risk, etc.
- 2. Adjust length of training period and number of clusters
- 3. Change/extend pool of stocks
- 4. Implement distance thresholds
- 5. Improve on the risk-return of the portfolio by: assigning weights on each stock of the portfolio based on an optimization problem to maximize Sharpe ratio
- 6. Add/utilize industry sector information for each company
- 7. Use time series analysis to describe the changes in stock prices throughout the duration

THANK YOU

Q&A