

Investment Theory & Practice—Asset Management
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Sorting by Trading Volume

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Investment Strategy

We are going to create a portfolio consisting of:

- 200 stocks with highest monthly trading volume
- Equally weighted
- Rebalance monthly:
 - Sell previous securities
 - Buy next month's securities

Rationale

- Higher trading volume stocks will likely have much higher buying trends versus low volume stocks -> higher buying trends, higher returns.
- Efficient market: Market participants will be more attracted to good fundamental companies so when there is a good earnings/good news -> more shares are purchased.
- The higher the volume, the tighter the Bid-Ask spread -> less cost incurred.

Data Pre-Processing

- **Data Acquisition:**

- We pulled all historical performances (with volume) from CRSP database for date range 01/01/1985 - 12/31/2019

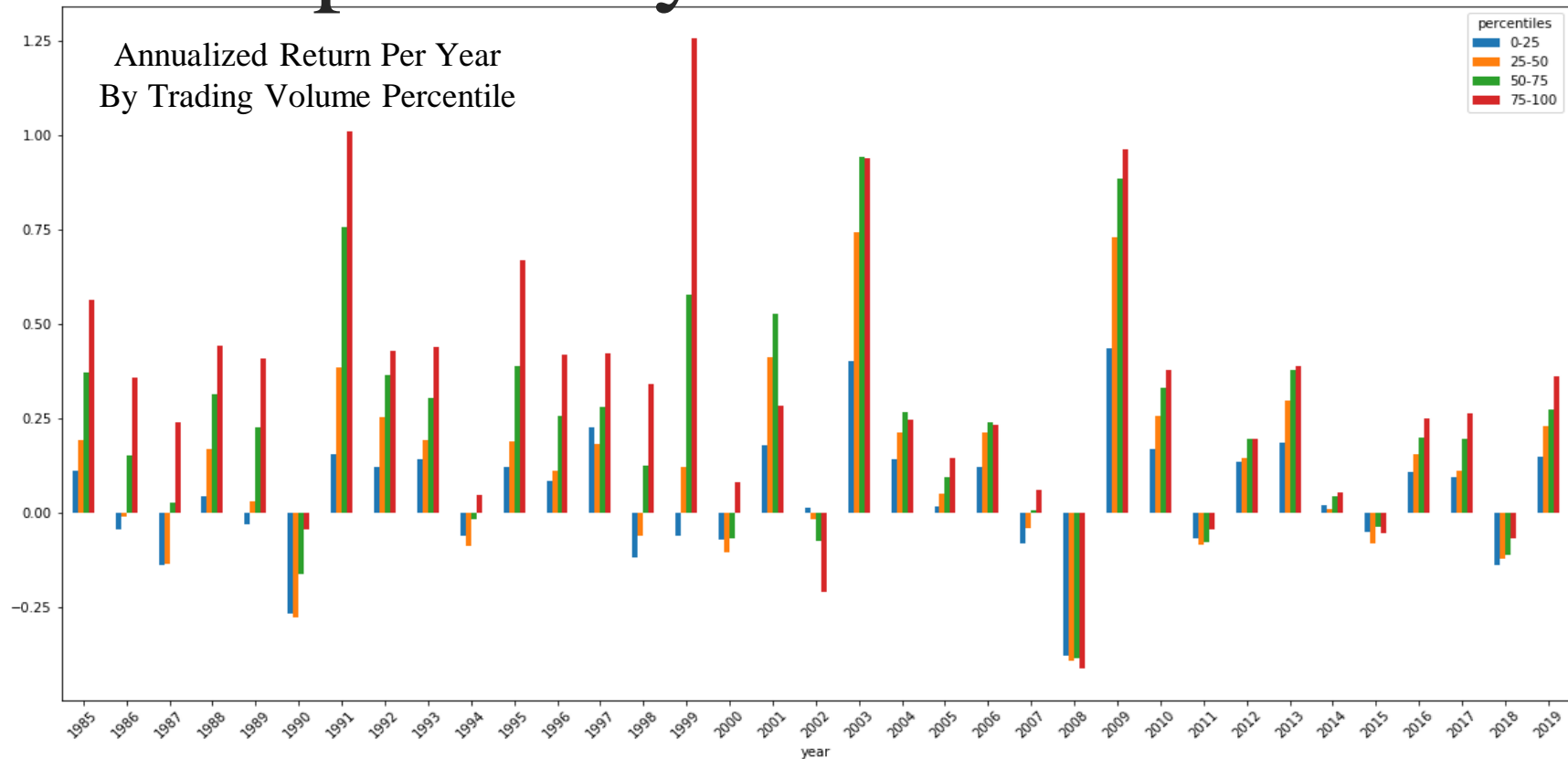
- **Data Cleanup:**

- We created a calculated field called “relative trading volume” by taking the monthly trading volume of a stock and divide it by the total trading volume of all stocks for that month.
- Eliminated the bottom 10% of firms by market cap because those volumes were ~ 0 .
- Split stocks to quartiles based on trading volume.

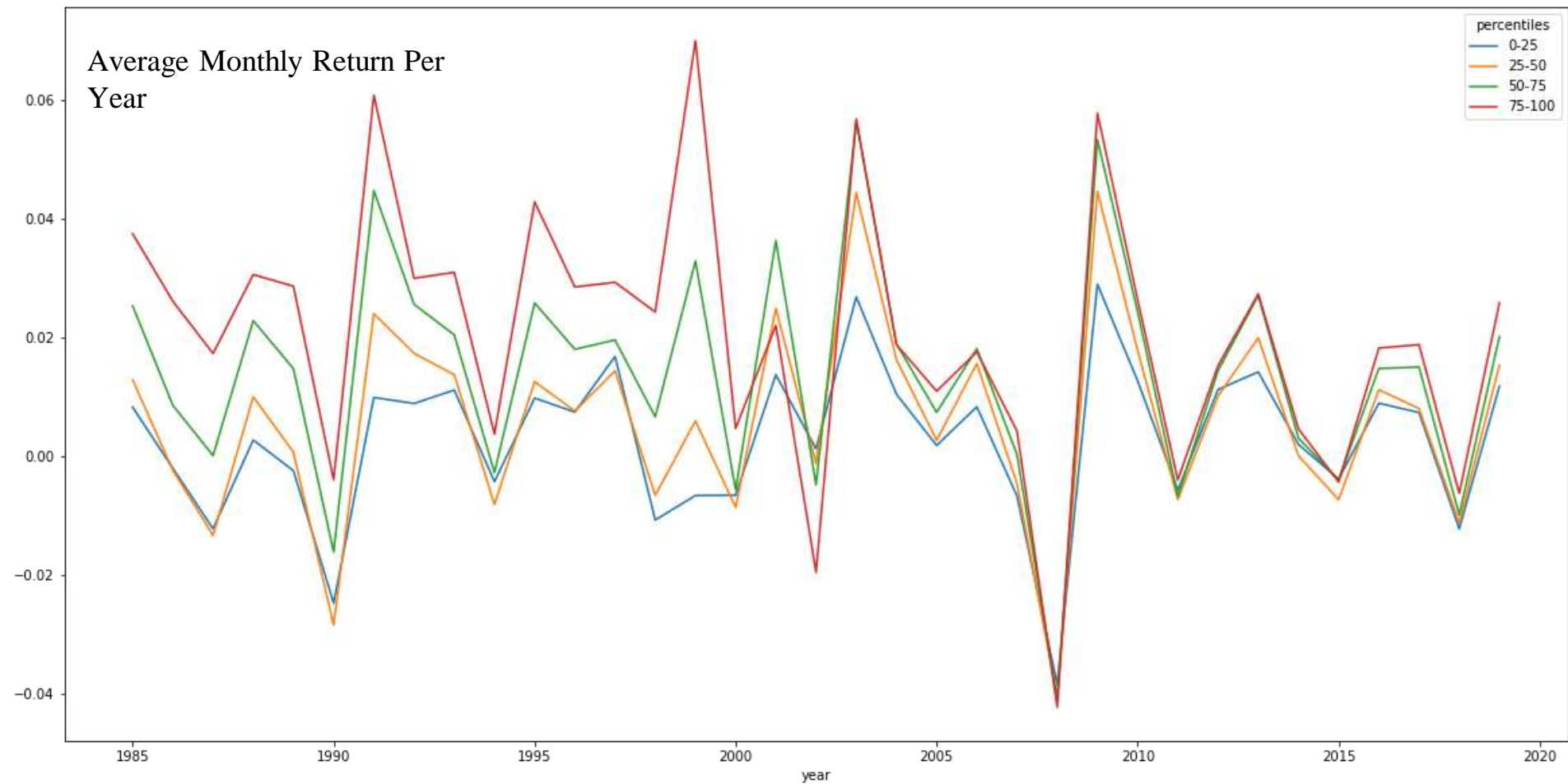
Findings

- High volume quartiles tends to outperform in favorable times.
- But in some time generate more losses in market downturns.

Exploratory & Statistical Proof



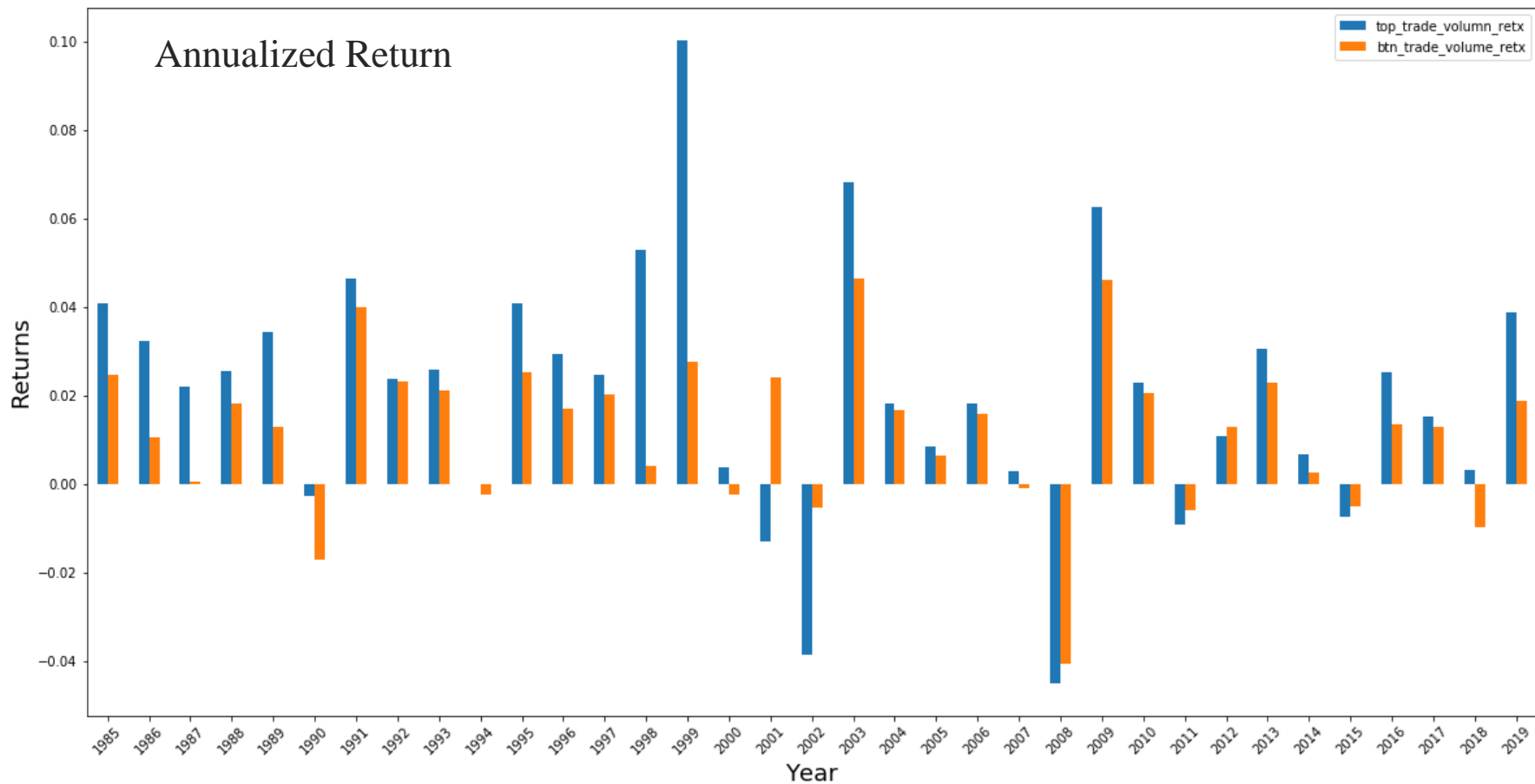
Average Monthly Return Per Year

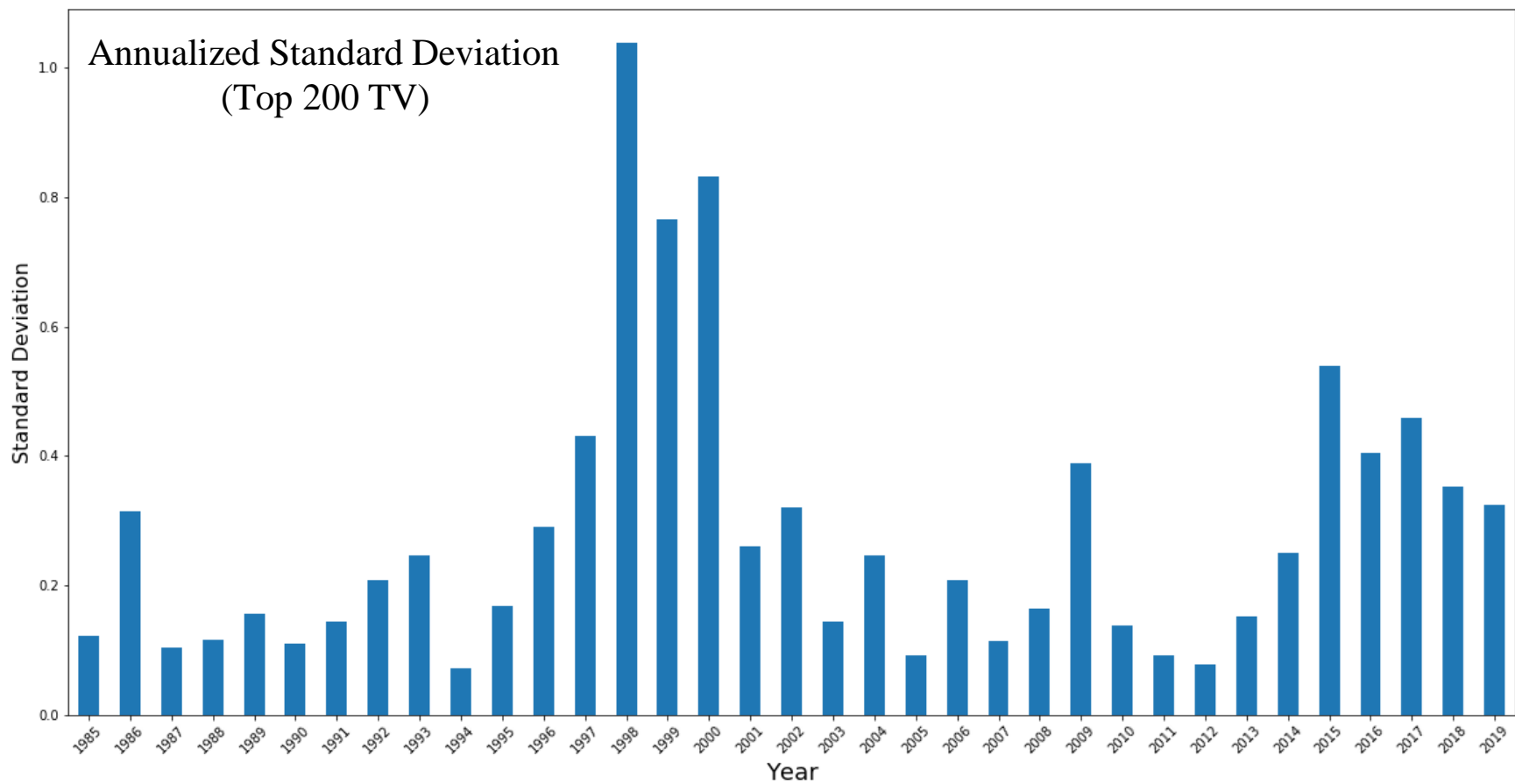


Portfolio Backtesting

We created 2 portfolios that consist of the top 200 and the bottom 200 relative trading volume stocks

Annualized Return





H-Null: The return of the 200 stock with highest trading volume is not statistically significant different from the 200 lowest trading volume stock

```
((diff['top_trade_volumn_retx'] - diff['btn_trade_volume_retx']).mean()) / ((diff['top_trade_volumn_retx'] -  
diff['btn_trade_volume_retx']).std()/math.sqrt(len(diff['top_trade_volumn_retx'])))
```

4.25, Reject !

H-Null: The return of top 25% trading volume is not statistically significant different from lowest 25% trading volume stocks

```
((percentile_retx['75-100'] - percentile_retx['0-25']).mean() - 0) / ((percentile_retx['75-100'] -  
percentile_retx['0-25']).std()/math.sqrt(len(percentile_retx['75-100'])))
```

7.686, Reject!

Final Results

- Assumptions:
- high volume portfolios trading cost of 5 BP
- low volume portfolios trading cost of 10 BP

	Top_200 Portfolio	Btm_200 Portfolio	Difference
Cost of Trading	5bp * 12 months	10 bp * 12 months	-
Avg Annual Return	27.711%	15.29%	12.421%
Avg Net Return	27.111%	14.29%	12.821%
Std (Annualized) (DrawDown %)	27.24%	18.01%	

Shortcomings

- Strategy assumes pre-knowledge of stocks with highest trading volume for that month.
- Strategy does not hedge against selloffs of low performing stocks in portfolio for each month.

How To Improve

- Research the relationship of high trading volume and momentum factor for better stock selection
- Hedge against low performing stocks by dropping security if it falls below mean return for all stocks in portfolio
- Add a factor in addition to volume to rebalance each month
Eg. Price to book ratio
- Volume alone does not factor in momentum 1000 shares of AMZN vs 1000 shares of BAC)
- Build a Predictive model to identify most traded volume stocks. Use relative volume for allocation instead of equally weighted.