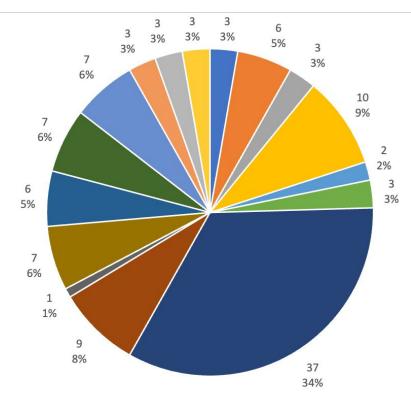
## DDMIF\_Final: Strategy Review

Group: Uhcakip Members: Ziyu Tang, Quan Cao, Baoyi Zhu, Tong Chen, Jiewan Yang

### Framework:

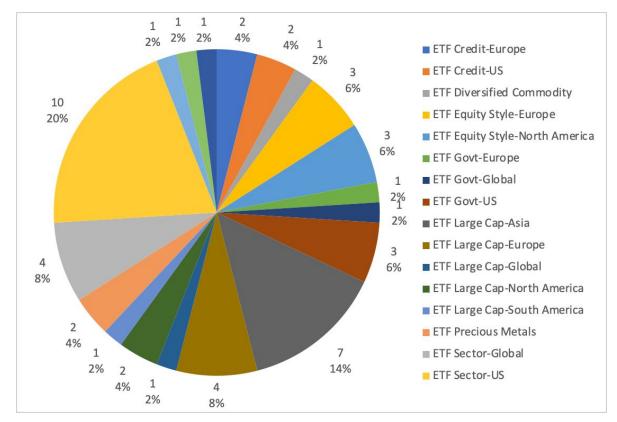
- 1. Introduction
- 2. Overall portfolio performance
- 3. Breakdown into our strategies
- 4. Fun parts of our portfolio
- 5. Summary

#### Universe - 110 assets

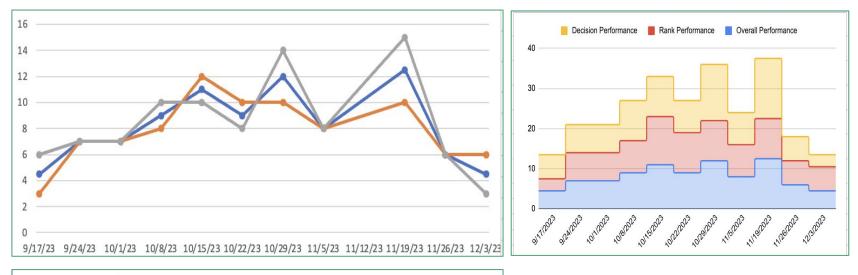




#### ETF Equities - 45% of Universe



#### II. Overall Portfolios Performance



	9/17/23	9/24/23	10/1/23	10/8/23	10/15/23	10/22/23	10/29/23	11/5/23	11/19/23	11/26/23	12/3/23
Overall Performance	4.5	7	7	9	11	9	12	8	12.5	6	4.5
Rank Performance	3	7	7	8	12	10	10	8	10	6	6
Decision Performance	6	7	7	10	10	8	14	8	15	6	3

#### Test this week's strategy by using last week's data:

	group_name	forecast_performance	decision_performance	forecasts_rank	decisions_rank	overall_rank
0	SP500	0.109976	8.144776	2.0	5.0	3.5
1	EW_LONG	0.16	10.644184	6.0	2.0	4.0
2	UHCAKIP	0.16	10.124099	6.0	3.0	4.5
3	GAMBLING	0.106469	5.172579	1.0	8.0	4.5
4	4SIGMA	0.12	5.2772	3.0	7.0	5.0
5	HIREUS	0.166429	12.467355	10.0	1.0	5.5
6	JAYSTREET	0.152587	-4.065978	4.0	12.0	8.0
7	DYF	0.186453	8.437688	12.0	4.0	8.0
8	MSG	0.161569	4.805087	7.0	9.0	8.0
9	HELLO_WORLD	0.173731	5.759857	11.0	6.0	8.5
10	ALPHA	0.152688	-5.128328	5.0	13.0	9.0
11	RANDOM	0.165008	2.517779	9.0	10.0	9.5
12	LAMM	0.16159	0.002227	8.0	11.0	9.5

#### III. Portfolios Structure

Number	Portfolio due	Strategy applied to ranking	Strategy applied to decisions	Overall performance
1	9/17/2023	Random	Random	4.5
2	9/24/2023	Random	Normal distribution	7.0
3	<u>10/1/2023</u>	Random	Linear Regression	7.0
4	<u>10/8/2023</u>	Random	Technical+Fundam ental	9.0

#### III. Portfolios Structure

Number	Portfolio due	Strategy applied to ranking	Strategy applied to decisions	Overall performance
5	<u>10/15/2023</u>	Technical factors	Turnover reverse	11.0
6	10/22/2023	Random	Volatility reverse	9.0
7	<u>10/29/2023</u>	Decision-Based	Subjective selection	12.0
8	<u>11/5/2023</u>	Random	Momentum	8.0

#### III. Portfolios Structure

Number	Portfolio due	Strategy applied to <b>ranking</b>	Strategy applied to <b>decisions</b>	Overall performance
9	<u>11/19/2023</u>	Technical factors	MVE portfolio	12.5
10	<u>11/26/2023</u>	Fundamental + optimization	Machine Learning	6
11	<u>12/3/2023</u>	Equal weights	Risk Parity	4.5

#### Oct 1 Ridge Linear Regression - Why Use

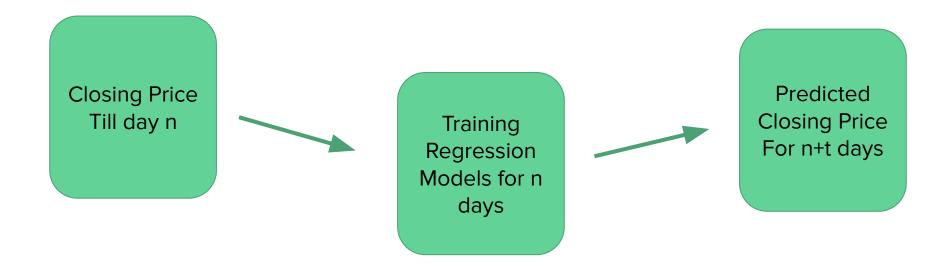
Linear Regression:

Easily be unstable when coefficients are getting larger.

Ridge Linear Regression:

Avoid the problem of predictor's collinearity by adding additional cost

#### Oct 1 Ridge Linear Regression - Framework 1



#### Oct 1 Ridge Linear Regression

L2 Penalty:

- Penalize the model based on the sum of the squared coefficient values  $\beta$
- Minimize the size of coefficients
- Lambda Penalty controls the weighting of the penalty

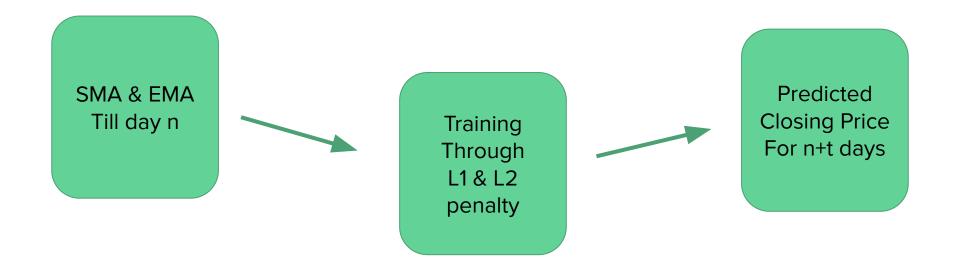
Indicators:

- Simple Moving Average
- Exponential Moving Average

MSE:

• To validate the prediction ability

#### Oct 1 Ridge Linear Regression - Framework 2



#### <u>Result</u>

	TICKER	pred_price
0	ABBV	139.130576
1	ACN	325.282652
2	AEP	81.494620
3	AIZ	142.944838
4	ALLE	105.665346
5	AMAT	144.422381
6	AMP	344.692232
7	AMZN	141.269001
8	AVB	171.275693
9	AVY	188.832350

Weighted decision after applying different methodology to technical and fundamental data

- Technical data: calculate beta value for decision
- Fundamental data
  - Select multiple fundamental attributes of a company
  - Use random forest regressor calculate predicted price
  - Use difference between predicted price and actual price to calculate decision
- Assign 0.5:0.5 weight to the different decisions

#### Nov 5 Strategy: Momentum

Calculate price returns for each stock over previous week

Set the benchmark return (s&p 500)

Calculate momentum score: individual return - benchmark return

Select positive momentum score stocks

• Standardize it to ensure sum of the decision = 1

#### Oct 15 & Oct 22 Reverse effect

- Calculated the excess turnover/volatility of each stock.
- Set a threshold and compared it with the excess turnover/volatility of each stock.
- For those stocks whose excess turnover/volatility lesses the threshold, stay the same; for those stocks whose excess turnover/volatility surpasses the threshold, we reverse its sign by multiplying -1.
- Standardization
- The idea aims to exploit potential market reversals or anomalies that deviate from the usual expectations.
- For the ranking, we chose RSI to balance the reverse effect.

$$RSI_{ ext{step one}} = 100 - \left[rac{100}{1 + rac{ ext{Average gain}}{ ext{Average loss}}}
ight]$$

#### Nov 19 MVE + Tech

- Calculated the covariance matrix.
- Maximize the expected return for a given level of risk or minimize risk for a specified level of expected return.
- Plot the efficient frontier, which represents the set of optimal portfolios.
- Identify the tangency portfolio.

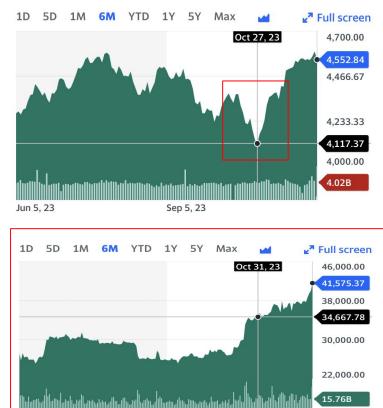
• For the rank, we combined RSI (risk strength index) with random strategy.

Overall, not working well...

• Assumption didn't hold, etc.

#### Oct 29 Strategy: Subjective Selection

- Market reaches phasal low, lack of market confidence;
- **Earnings Release**: Subjectively selected 5 stocks with earnings surprise (reported earnings more than consensus estimates).
- **Bitcoin** in upward momentum, **Gold** is an alternative for downward market pressure
- **Short** the rest of the investment universe with equal weights.

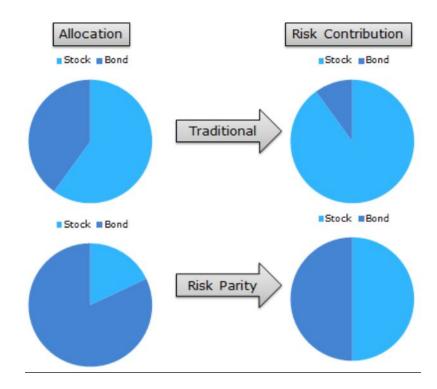


Sep 4, 23

Jun 4, 23

#### Dec 3 Strategy: Risk Parity

- Traditional allocation: 60% Equity, 40%
   Fixed Income. Inequivalent Risk
   Contribution.
- Risk Parity: capital allocation to ensure equal risk contribution for each security.
- Benefit: safter during recession, risk diversification;
- Limitation: market timing, increased cash allocation, sensitive to correlation



#### Dec 3 Strategy: Risk Parity

- Data used: closing price for every ticker from 2023-01-01 to 2023-12-01, a daily log-return is calculated based on closing prices.
- Risk: Portfolio Volatility is determined by the return covariance matrix and weights.

- Marginal Risk Contribution: Dot product of weights and covariance matrix divided by the portfolio volatility (percentage of risk contribution)
- Target: equal risk contribution across all tickers

$$\sigma_p^2 = W^t * Covariance Matrix * W$$

	ABBV	ACN	ADA-USD	AEP	AIZ	ALLE
ABBV	0.000163	0.000007	-0.000022	0.000040	0.000035	-0.000004
ACN	0.000007	0.000202	0.000108	0.000034	0.000062	0.000112
ADA- USD	-0.000022	0.000108	0.001572	0.000036	0.000009	0.000115
AEP	0.000040	0.000034	0.000036	0.000175	0.000037	0.000047
AIZ	0.000035	0.000062	0.000009	0.000037	0.000315	0.000083



	Methodology					
Ranking:		Fundamental factors	Optimization threshold			
Rank 1 Rank 2 Rank 3		Price to equity ratio	25			
	Optimization +	Price to cash flow	10			
	fundamental factor	Price to book value	0.8			
Rank 4		Debt to equity ratio	1.5			
Rank 5		Price to sales ratio	1.5			
Decision	Random forecast: pre	edict price -> calculate the pe weight	ercentage change ->			

# III. Reverse

Uhcakip

Pikachu

	group_name	mean_forecast	mean_decision	forecasts_rank	decisions_rank	overall_rank
0	EW_LONG	0.160000	6.119650	4.0	2.0	3.0
1	SP500	0.133970	3.027431	1.0	8.0	4.5
2	HIREUS	0.162895	3.629035	5.0	6.0	5.5
3	GAMBLING	0.144686	2.705927	2.0	10.0	6.0
4	UHCAKIPCT	0.183835	10.124099	13.0	1.0	7.0
5	UHCAKIPZIYU	0.174224	4.460735	11.0	5.0	8.0
6	HELLO_WORLD	0.184591	5.399520	14.0	3.0	8.5
7	JAYSTREET	0.158699	-5.097778	3.0	17.0	10.0
8	4SIGMA	0.217930	4.697089	16.0	4.0	10.0
9	RANDOM	0.172858	0.818478	10.0	12.0	11.0
10	MEDALLION	0.171858	-0.677937	8.0	15.0	11.5
11	PIKACHU	0.170917	-0.737002	7.0	16.0	11.5
12	ALPHA	0.188541	2.831275	15.0	9.0	12.0
13	UHCAKIPBY	0.166182	-8.055181	6.0	19.0	12.5
14	LAMM	0.182170	0.216732	12.0	13.0	12.5
15	CITADELSPINOFFS	0.236956	3.060047	19.0	7.0	13.0
16	UHCAKIP	0.171929	-6.304314	9.0	18.0	13.5
17	DYF	0.221980	2.466174	17.0	11.0	14.0
18	MSG	0.222321	-0.437691	18.0	14.0	16.0

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#### IV. Summary

- Don't believe the theoretical knowledge
- Shorting is very risky
- Hedging the risk is necessary
- Timing is important
  - Subjective selection case