
Predicting Volatility Regimes in Coinbase Stock

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Motivation

Bitcoin is considered a **safe-haven asset**,
largely uncorrelated from stock markets

**Could there be predictive power in Bitcoin
price movements?**

Goal

Create a model that uses **Bitcoin** price movements to predict near future **Coinbase volatility** spikes

Why Coinbase?

- We picked Coinbase because we'd traded it and kept seeing sharp reactions to market sentiment.
- Coinbase earns revenue from trading fees, so volatility should help them, yet the stock mainly moves with BTC's price level, making that disconnect worth studying.
- Coinbase stock behaves like a leveraged play on Bitcoin, making volatility the key thing to predict.
- Bitcoin often moves first, so we tested whether BTC signals can anticipate COIN's biggest swings.

Who is this for?

Quant firms **already** have models and low-latency infrastructure that we would never be able to come close to

Meanwhile, **non-quant investors** could easily benefit from a short-term volatility indicator for many types of trading strategies, such as a straddle options strategy

Modeling

Prediction Target

Predicting BINARY outcome: "Is the next period a volatile window?"

Definition: **volatile window** is determined true if next set of move is greater than the rolling median of the last 50 periods.

Feature Selection

Despite using Bitcoin as the main signal source, its **high volatility** prompted us to look for **secondary signals** that would **shield our model** from being too swayed by bigger swings in Bitcoin price

Features tested:

- Economic factors: HML, SMB, Beta, Momentum
- Mega-cap stocks (AAPL, MSFT, NVDA)
- Market benchmarks: S&P, Russell
- Volatility index: VIX

Dealing with look-ahead bias

- Used **time-ordered train, validation, and test splits**
 - Ensuring the model never trained on future information.
- Tuned **hyperparameters** only on validation window.
 - Final evaluation reserved for a fully out-of-sample test period.
- Mirrors **real world** deployment conditions.
 - Model is evaluated on data that comes after training.

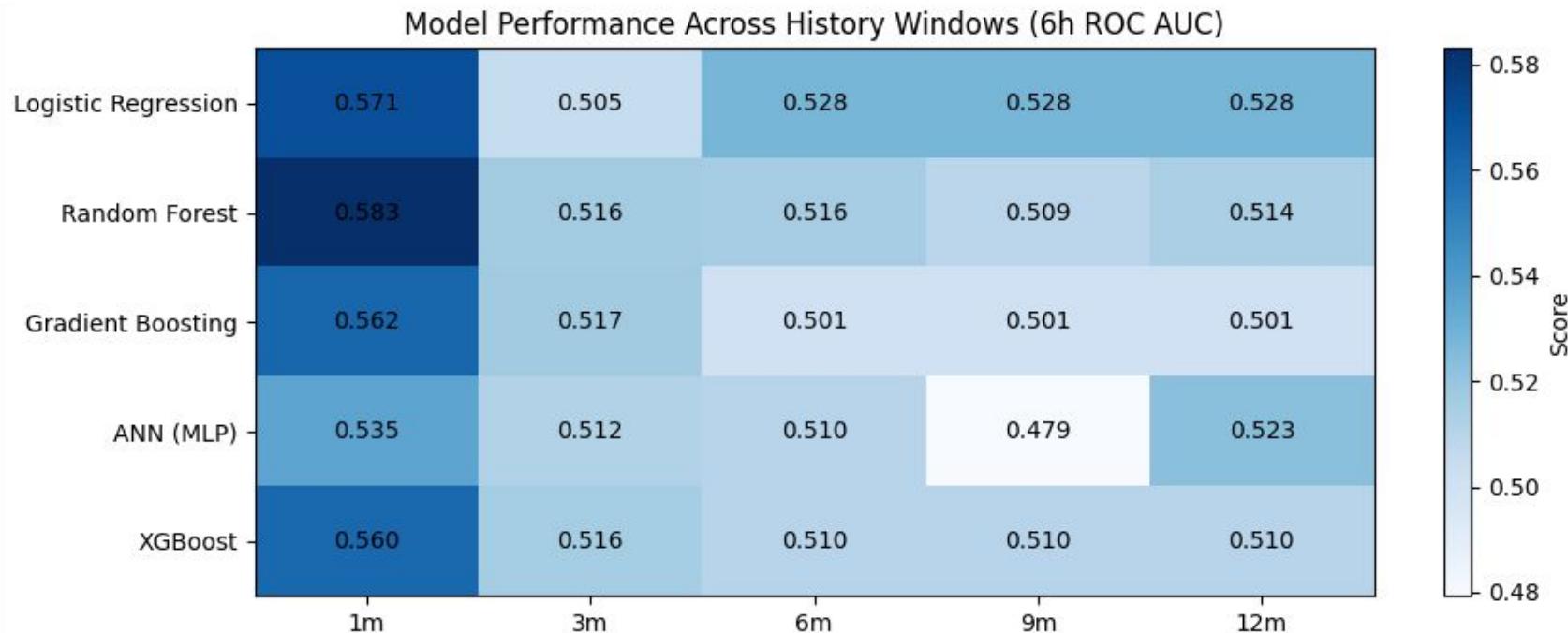
Training Time Frame

High volatility of BTC encourages a **shorter** training time frame

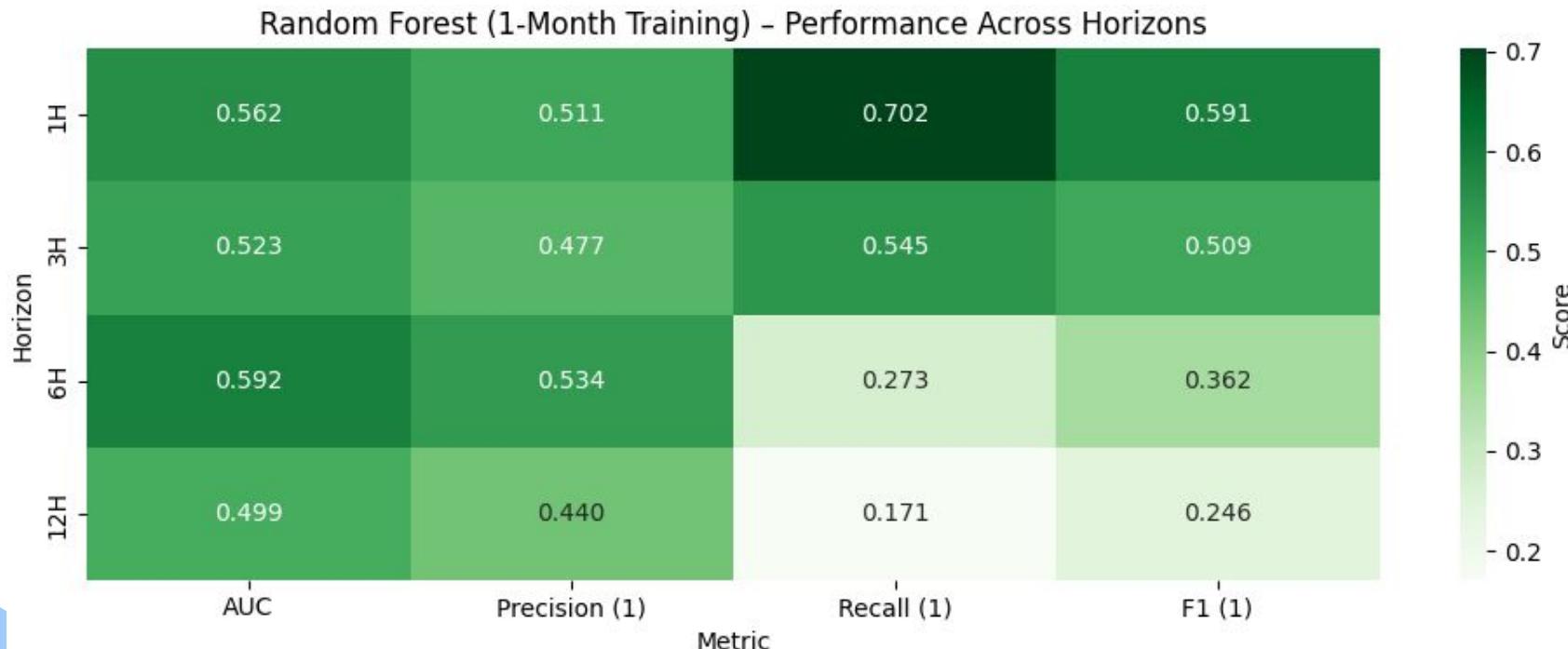
Outdated patterns won't affect predictability of current trends...

... but, too **narrow** windows will very easily lead to **overfitting**

Models Tested vs Training Windows



RF Model Results / Prediction Horizons



Model Interpretation

Prediction Target: **Volatility**

Precision: Shows reliable signal

51%

- Model says "large move" incoming => how true?
=> filters out half of the false alarms => less unnecessary losses.

Recall: Shows enough signal to be reliable

70%

- Of ALL real "large moves", how often does the model identify them?
=> identifies half the volatile windows => potential "opportunity" windows;
trade or not, is up to individual at this point.

F1: Both.

59%

- F1 shows when price swings happen (Recall), and that it's reliable (Precision).

AUC-ROC: Model indeed has signal.

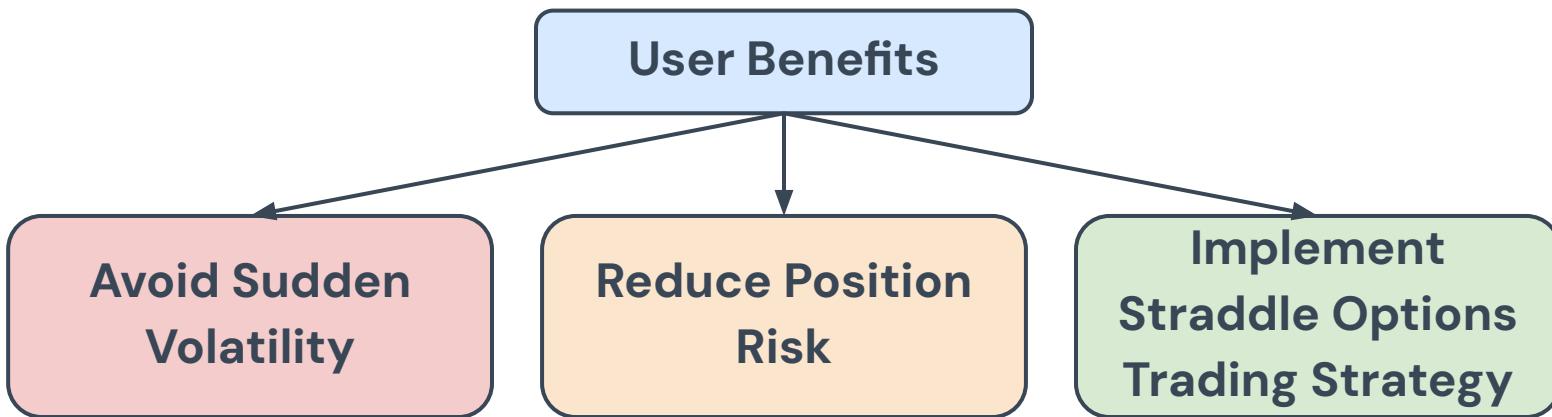
56%

SET ASIDE: **Accuracy**

*more models in appendix

Business Takeaway

Our model predicts large price swings over a 1-hour window.



Implementation

Platform Integration



Independent Usage

- Mobile App
- Browser Extension
- TradingView Indicator

Thanks!

Any questions?

Appendix

Sample size: 1731
Train: 969 | Val: 242 | Test: 520
Class balance in full sample (0=no large move, 1=large move):
class 0: 911 (52.6%)
class 1: 820 (47.4%)

1H Horizon

XGBoost (Best tuned on val, refit on Train+Val)

Accuracy: 54.038%

ROC AUC: 0.5362

Confusion Matrix:

```
[[142 127]
 [112 139]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.559	0.528	0.543	269
1	0.523	0.554	0.538	251
accuracy			0.540	520
macro avg	0.541	0.541	0.540	520
weighted avg	0.541	0.540	0.540	520

XGBoost Feature Importances:

```
coin_ret_1h      0.112594
coin_minus_btc  0.108686
btc_vol_6h      0.104581
btc_ret_lag_1   0.101507
coin_ret_lag_3   0.098962
coin_ret_lag_2   0.097359
coin_ret_lag_1   0.097044
coin_vol_6h     0.095198
btc_ret_lag_2   0.093845
btc_ret_1h      0.090223
```

3H Horizon

XGBoost (Best tuned on val, refit on Train+Val)

Accuracy: 52.115%

ROC AUC: 0.5458

Confusion Matrix:

```
[[137 114]
 [135 134]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.504	0.546	0.524	251
1	0.540	0.498	0.518	269
accuracy			0.521	520
macro avg	0.522	0.522	0.521	520
weighted avg	0.523	0.521	0.521	520

XGBoost Feature Importances:

```
coin_ret_1h      0.108898
coin_minus_btc  0.104527
btc_vol_6h      0.101893
coin_ret_lag_2   0.101026
btc_ret_lag_1   0.100867
btc_ret_lag_2   0.100061
coin_minus_btc  0.097254
coin_ret_lag_3   0.097031
btc_ret_1h      0.095638
coin_ret_lag_1   0.092805
```

6H Horizon

XGBoost (Best tuned on val, refit on Train+Val)

Accuracy: 53.077%

ROC AUC: 0.5490

Confusion Matrix:

```
[[149 108]
 [136 127]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.523	0.580	0.550	257
1	0.540	0.483	0.510	263
accuracy			0.531	520
macro avg	0.532	0.531	0.530	520
weighted avg	0.532	0.531	0.530	520

XGBoost Feature Importances:

```
btc_vol_6h      0.112538
btc_ret_1h      0.112536
coin_ret_lag_2   0.107217
coin_vol_6h      0.105637
coin_ret_1h      0.101768
btc_ret_lag_2    0.097017
btc_ret_lag_1    0.094319
coin_ret_lag_1   0.090958
coin_ret_lag_3   0.089689
coin_minus_btc  0.088322
```

12H Horizon

XGBoost (Best tuned on val, refit on Train+Val)

Accuracy: 52.692%

ROC AUC: 0.5410

Confusion Matrix:

```
[[157 105]
 [141 117]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.527	0.599	0.561	262
1	0.527	0.453	0.487	258
accuracy			0.527	520
macro avg	0.527	0.526	0.524	520
weighted avg	0.527	0.527	0.524	520

XGBoost Feature Importances:

```
coin_vol_6h      0.112660
btc_vol_6h      0.112189
coin_ret_lag_3   0.100470
btc_ret_1h      0.100392
coin_minus_btc  0.100156
btc_ret_lag_2    0.099230
coin_ret_lag_2   0.098082
btc_ret_lag_1    0.093292
coin_ret_lag_1   0.091793
coin_ret_1h      0.091737
```

24H Horizon

```
XGBoost (Best tuned on val, refit on Train+Val)
```

```
Accuracy: 47.308%
```

```
ROC AUC: 0.4640
```

```
Confusion Matrix:
```

```
[[131 151]
 [123 115]]
```

```
Classification Report:
```

	precision	recall	f1-score	support
0	0.516	0.465	0.489	282
1	0.432	0.483	0.456	238
accuracy			0.473	520
macro avg	0.474	0.474	0.473	520
weighted avg	0.478	0.473	0.474	520

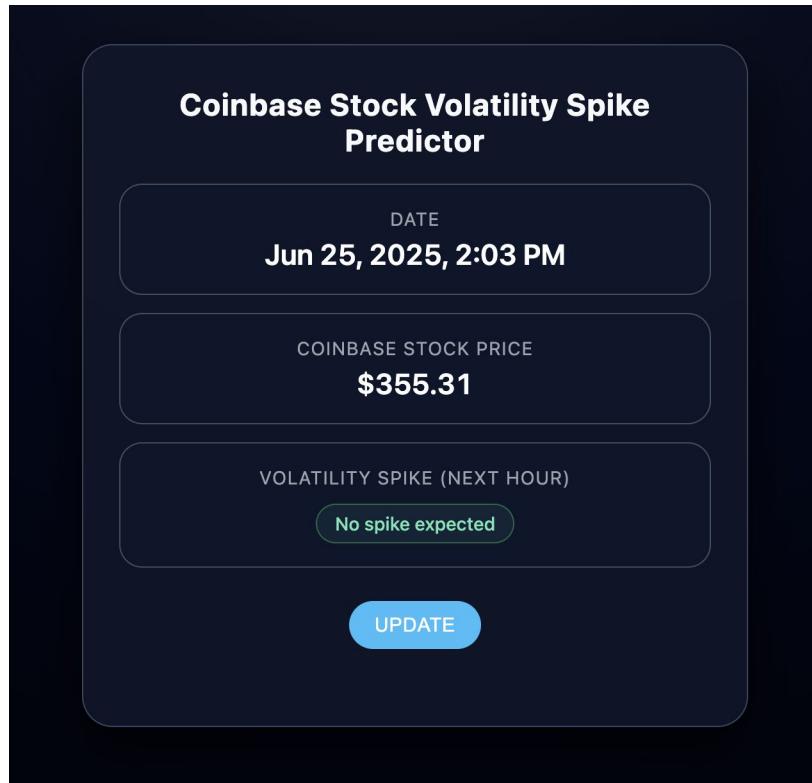
```
XGBoost Feature Importances:
```

```
coin_vol_6h      0.113643
btc_vol_6h       0.110631
coin_minus_btc   0.101362
btc_ret_lag_1     0.098599
coin_ret_1h       0.097747
btc_ret_lag_2     0.097670
coin_ret_lag_2    0.097330
btc_ret_1h        0.096680
coin_ret_lag_1    0.095891
coin_ret_lag_3    0.090447
```

Model Results (RF 6H)

Horizon	AUC	Prec (1)	Recall (1)	F1 (1)
1H	0.563	0.521	0.628	0.570
3H	0.498	0.451	0.489	0.469
6H	0.579	0.547	0.489	0.516
12H	0.529	0.472	0.389	0.426

API Demo Screenshot



API Demo Repo

<https://github.com/fe-vianna/ML-Final-Project>