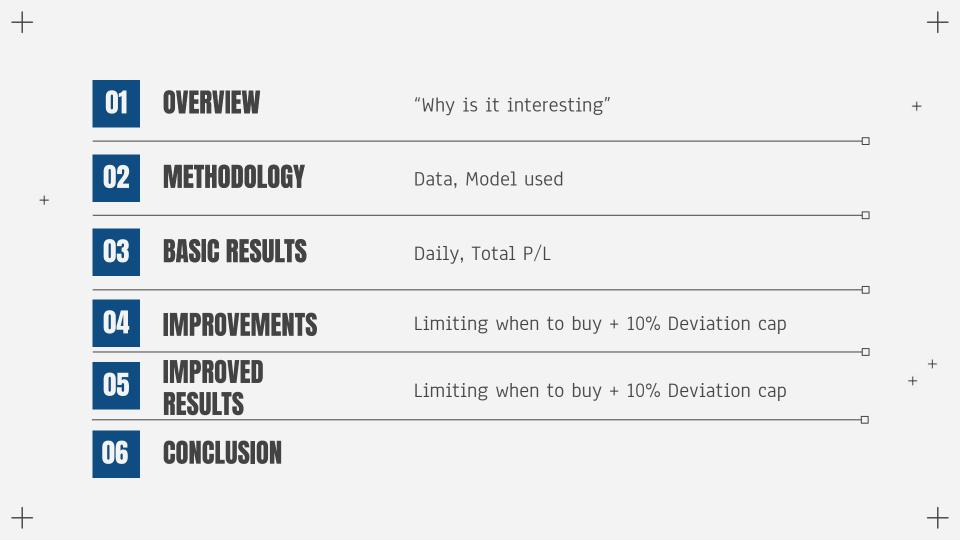
SPX vs SPY

OPTIONS PRICING PROJECT +

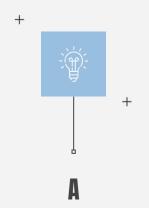
QF101 G1T3
JUAN, ENJIE, DOMINIQUE, IAN



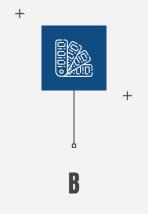


+

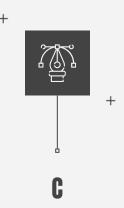
ABOUT THE PROJECT



Based on: S&P 500 SPX - European SPY - American



SPY is the most traded option



Large difference in trade volume SPX: 1-3 Million daily SPY: 60-140 Million daily

Overview - Steps

- 1. Create our own Options Pricing Model
- 2. Compare the predicted option prices against actual market prices of the options
- 3. Compare how our model does in predicting SPX options (European) vs SPY options (American)

Improvement:

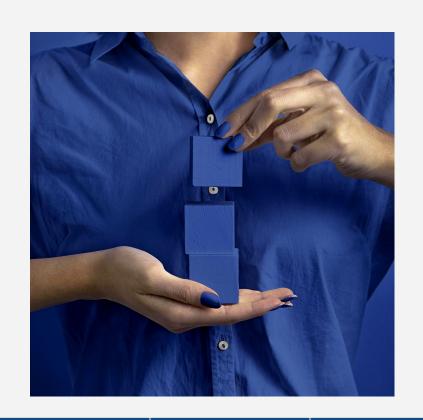
Go long when our predicted value is more than 0% to 10% of the market value, Go short when our predicted value is less than 0% to 10% of the market value





Data range: 3/1/2020 to 30/12/2022

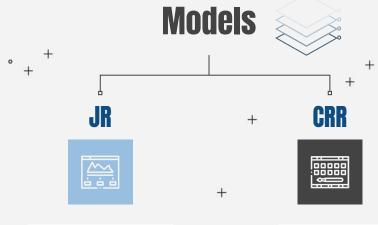
Options: 30 Days to Expiry



+

Code

```
# calculate the risk-neutral probability
if method == 'crr':
    u = np.exp(sigma * np.sqrt(dt))
    d = 1 / u
    p = (np.exp((r - div) * dt) - d) / (u - d)
elif method == 'jr':
    u = np.exp((r-div-0.5*sigma**2)*dt+sigma*np.sqrt(dt))
    d = np.exp((r-div-0.5*sigma**2)*dt-sigma*np.sqrt(dt))
    p = 0.5
```



$$p = \frac{1}{2}$$

$$u = e^{\sigma \sqrt{dt}}$$

$$u = e^{(r - \frac{1}{2}\sigma^2)dt + \sigma \sqrt{dt}}$$

$$d = e^{(r - \frac{1}{2}\sigma^2)dt - \sigma \sqrt{dt}}$$

$$d = e^{(r - \frac{1}{2}\sigma^2)dt - \sigma \sqrt{dt}}$$

$$p = \frac{e^{rdt} - d}{u - d}$$

Buy Signals

Long Call/Put

Enter +1 position when predicted price > market price.

Short Call/Put

Enter -1 position when predicted price < market price.

```
df_input['LONG_CALL'] = np.where(df_input['CALL'] > df_input['C_PRICE'], long_call, short_call)
df_input['LONG_PUT'] = np.where(df_input['PUT'] > df_input['P_PRICE'], long_put, short_put)
dcall = (df_input['CALL'] - df_input['C_PRICE'])/df_input['C_PRICE']
dput = (df_input['PUT'] - df_input['P_PRICE'])/df_input['P_PRICE']
```

Calculating PnL

- 1. Calculate cost of buying / premium received from entering into a position (PROFIT LOSS ON PURCHASE)
- 2. Calculate options payoff (PROFIT_LOSS_ON_SALE):
 - a. Call: Max(Stock Price Strike Price, 0)
 - b. Put: Max(Strike Price Stock Price, 0)
- 3. PnL = PROFIT LOSS ON PURCHASE + PROFIT LOSS ON SALE

```
df_input['PROFIT_LOSS_ON_PURCHASE_CALL'] = -(df_input['LONG_CALL']*df_input['C_PRICE'])
df_input['PROFIT_LOSS_ON_PURCHASE_PUT'] = -(df_input['LONG_PUT']*df_input['P_PRICE'])

df_input['PROFIT_LOSS_ON_PURCHASE'] = df_input['PROFIT_LOSS_ON_PURCHASE_CALL'] + df_input['PROFIT_LOSS_ON_PURCHASE_PUT']

df_input['PROFIT_LOSS_ON_SALE_CALL'] = (df_input['LONG_CALL']*np.maximum(df_input['UNDERLYING_PRICE_EXPIRE'] - df_input['STRIKE'], 0))

df_input['PROFIT_LOSS_ON_SALE_PUT'] = (df_input['LONG_PUT']*np.maximum(df_input['STRIKE'] - df_input['UNDERLYING_PRICE_EXPIRE'], 0))

df_input['PROFIT_LOSS_ON_SALE'] = df_input['PROFIT_LOSS_ON_SALE_CALL'] + df_input['PROFIT_LOSS_ON_SALE_PUT']

df_input['PROFIT_LOSS_CALL'] = df_input['PROFIT_LOSS_ON_SALE_CALL'] + df_input['PROFIT_LOSS_ON_PURCHASE_CALL']

df_input['PROFIT_LOSS_PUT'] = df_input['PROFIT_LOSS_ON_SALE_PUT'] + df_input['PROFIT_LOSS_ON_PURCHASE_PUT']

df_input['PROFIT_LOSS'] = df_input['PROFIT_LOSS_ON_SALE'] + df_input['PROFIT_LOSS_ON_PURCHASE']
```



SPX

SPY

Root Mean Squared Error of CRR Call model: 20.894495881960978

Root Mean Squared Error of JR Call model: 20.892574928621944

Root Mean Squared Error of CRR Put model: 21.61877256944796

Root Mean Squared Error of JR Put model: 21.6169893240666

Mean Absolute Error of CRR Call model: 12.179502242149622

Mean Absolute Error of JR Call model: 12.178586841592729

Mean Absolute Error of CRR Put model: 13.381022873010346

Mean Absolute Error of JR Put model: 13.3795543630229

Percentage Mean Absolute Error of CRR Call model: 89.90169561712081 %

Percentage Mean Absolute Error of JR Call model: 89.7479891380906 %

Percentage Mean Absolute Error of CRR Put model: 52.08037360182608 %

Percentage Mean Absolute Error of JR Put model: 52.073426146778615 %

Mean Profit/Loss of CRR model: 8.959887290942877
Mean Profit/Loss of JR model: 8.900853060784788

Volatility of CRR model: 207.2139025439408

[Volatility of JR model: 207.2486516839786]

Total Profit/Loss of CRR model: 332689.57499999995
Total Profit/Loss of JR model: 330497.57499999995

Root Mean Squared Error of CRR Call model: 2.2014083365619896
Root Mean Squared Error of JR Call model: 2.201059135659596
Root Mean Squared Error of CRR Put model: 2.221104470891091
Root Mean Squared Error of JR Put model: 2.2208519746157314
Mean Absolute Error of CRR Call model: 1.33057266941219
Mean Absolute Error of JR Call model: 1.3303482365965098
Mean Absolute Error of CRR Put model: 1.398773103360051
Mean Absolute Error of JR Put model: 1.3986049776617586
Percentage Mean Absolute Error of CRR Call model: inf %
Percentage Mean Absolute Error of JR Call model: inf %
Percentage Mean Absolute Error of CRR Put model: 46.934619008242365 %
Percentage Mean Absolute Error of JR Put model: 46.93023731003161 %

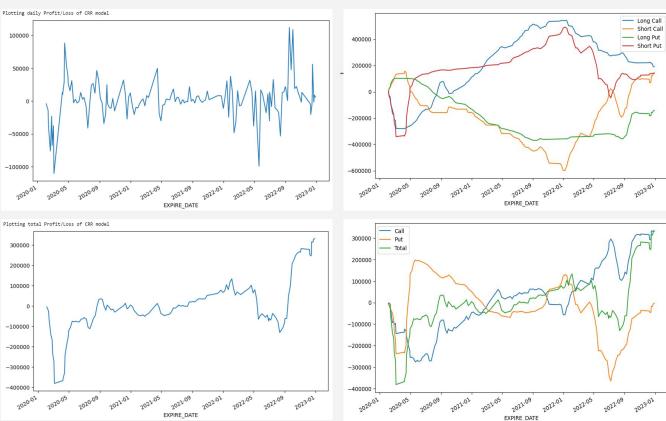
Mean Profit/Loss of CRR model: -0.5089830091975012
Mean Profit/Loss of JR model: -0.5032064873973185
Volatility of CRR model: 19.96222973690234
Volatility of JR model: 19.96649186704153
Total Profit/Loss of CRR model: -14498.89000000002

Total Profit/Loss of JR model: -14334.340000000015



Daily/Total

Call vs Put

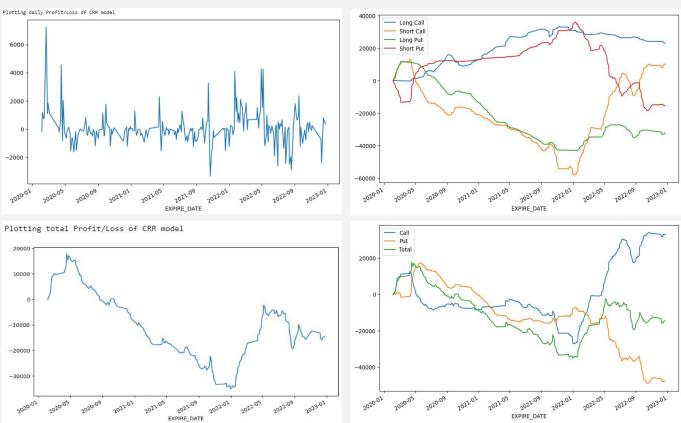


+



Daily/Total

Call vs Put



+

LIMITING WHEN TO BUY & 10% PRICE DEVIATION CONSTRAINT

```
# df_input['LONG CALL'] = np.where(df_input['CALL'] > df_input['C PRICE'], long call, short_call)
# df_input['LONG_PUT'] = np.where(df_input['PUT'] > df_input['P_PRICE'], long_put, short_put)
dcall = (df_input['CALL'] - df_input['C_PRICE'])/df_input['C_PRICE']
dput = (df input['PUT'] - df input['P PRICE'])/df input['P PRICE']
df input['LONG CALL'] = np.where(dcall > 0, np.where(dcall <= 0.1, long call, 0), np.where(dcall >= -0.1, short call, 0))
df input['LONG PUT'] = np.where(dput > 0, np.where(dput <= 0.1, long put, 0), np.where(dput >= -0.1, short_put, 0))
```



SPX

SPY

Root Mean Squared Error of CRR Call model: 20.894495881960978

Root Mean Squared Error of JR Call model: 20.892574928621944

Root Mean Squared Error of CRR Put model: 21.61877256944796

Root Mean Squared Error of JR Put model: 21.6169893240666

Mean Absolute Error of CRR Call model: 12.179502242149622

Mean Absolute Error of JR Call model: 12.178586841592729

Mean Absolute Error of CRR Put model: 13.381022873010346

Mean Absolute Error of JR Put model: 13.3795543630229

Percentage Mean Absolute Error of CRR Call model: 89.90169561712081 %

Percentage Mean Absolute Error of JR Call model: 89.7479891380906 %

Percentage Mean Absolute Error of CRR Put model: 52.08037360182608 %

Percentage Mean Absolute Error of JR Put model: 52.073426146778615 %

Root Mean Squared Error of CRR Call model: 2.2014083365619896
Root Mean Squared Error of JR Call model: 2.201059135659596
Root Mean Squared Error of CRR Put model: 2.22104470891091
Root Mean Squared Error of JR Put model: 2.2208519746157314
Mean Absolute Error of CRR Call model: 1.33057266941219
Mean Absolute Error of JR Call model: 1.3303482365065098
Mean Absolute Error of CRR Put model: 1.398773103360051
Mean Absolute Error of JR Put model: 1.3986049776617586
Percentage Mean Absolute Error of CRR Call model: inf %
Percentage Mean Absolute Error of JR Call model: inf %
Percentage Mean Absolute Error of CRR Put model: 46.934619008242365 %
Percentage Mean Absolute Error of JR Put model: 46.934619008242365 %
Percentage Mean Absolute Error of JR Put model: 46.93023731003161 %

Mean Profit/Loss of CRR model: 14.05162047884517

Mean Profit/Loss of JR model: 13.985484500821414

Volatility of CRR model: 198.01085248543916

Volatility of JR model: 198.03539355460376

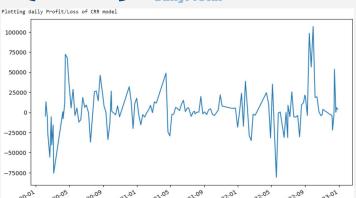
Total Profit/Loss of CRR model: 521750.72

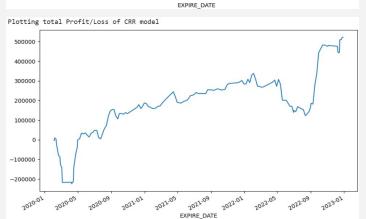
Total Profit/Loss of JR model: 519295.0249999999

Mean Profit/Loss of CRR model: 0.2437407849469913
Mean Profit/Loss of JR model: 0.25121621147230183
Volatility of CRR model: 19.076043535318924
Volatility of JR model: 19.080055467270356
Total Profit/Loss of CRR model: 6943.19999999994
Total Profit/Loss of JR model: 7156.14499999999

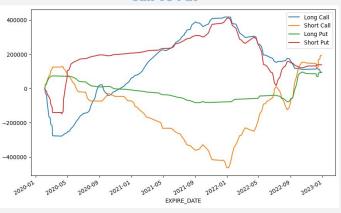


Daily/Total





Call vs Put







OVERVIEW

METHODOLOGY

RESULTS

IMPROVEMENTS

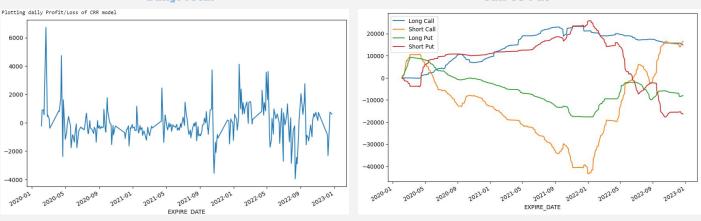
NEW RESULTS

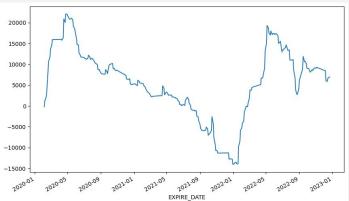
CONCLUSION

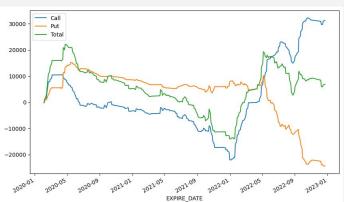
*SPY (P/L)

Daily/Total

Call vs Put







+

OVERVIEW

METHODOLOGY

RESULTS

IMPROVEMENTS

NEW RESULTS

CONCLUSION



Lessons Learnt

- SPX has better returns, probably due to the lower trade volume.
- Lower liquidity = Higher volatility = More gains :-)
- SPX = 10 * SPY
- CRR & JR gives similar results

° + ⁺

THANKS!



+