Options Prediction with Advanced DL

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Understanding Options Pricing

Lets watch a video first!

- Options are financial derivatives that gives the right to buy (call) or sell (put) a security at or before a certain date.
- We use deep learning to construct models that try to price.
- Options using historical data.

5 Basic Characteristics of Every Option

- Underlying asset
- Call vs. put
- Strike price
- Expiration date
- American vs. European

Why?

The value of a European option can be modeled with the commonly-used

Black-Scholes
Model (BS-Model):

$$C=N(d_1)S_t-N(d_2)Ke^{-rt}$$
 where $d_1=rac{\lnrac{S_t}{K}+(r+rac{\sigma^2}{2})t}{\sigma\sqrt{t}}$ and $d_2=d_1-\sigma\sqrt{t}$

C = call option price

N = CDF of the normal distribution

 S_t = spot price of an asset

 \boldsymbol{K} = strike price

r = risk-free interest rate

t = time to maturity

 σ = volatility of the asset

This model makes many <u>assumptions</u> (especially of volatility) and often mismatches empirical findings in different options.

How?

We used LSTM(model1) and MLP(model2,model3) architectures to predict the options prices and got results that outperform the Black-Scholes model significantly.

- 1. Data
- 2. Models
- 3. Results

Dataset

Stock data from yahoo finance

https://finance.yahoo.com/

• Option data from yahoo finance option

https://finance.yahoo.com/quote/WIMI/options?p=WIMI

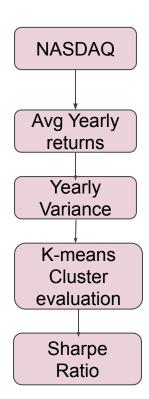
Treasury Rate

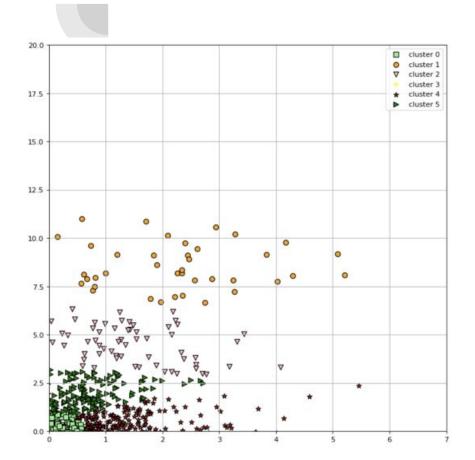
https://home.treasury.gov/

Stock Picking

How to choose the best stocks?

Strategy



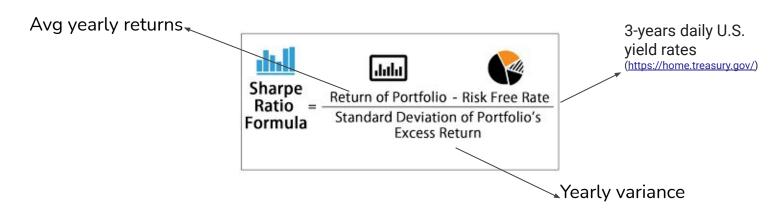


Applying K-means Clustering

Cluster '0' further reduced down to 6 clusters-

	cluster	avg_yearly_returns	yearly_variance	Name
0	3	-0.380375	0.096733	208
1	0	-0.007581	0.063594	984
2	5	0.105799	0.408778	204
3	2	0.284402	0.842208	75
4	4	0.359156	0.130867	246
5	1	0.560519	1.522883	40

Sharpe Ratio for evaluating the best Stocks Cluster



	cluster	avg_yearly_returns	yearly_variance	sharpe_ratio	Name
0	0	0.289101	0.067665	1.677841	137
1	1	0.326611	0.224425	0.649311	80
2	2	0.779880	0.171351	2.216106	29

Highest Sharpe Ratio

	Name	avg_yearly_returns	yearly_variance	cluster	avg_risk_free_rate	std_dev	sharpe_ratio
1105	NSTG	0.841255	0.022083	2	0.021367	0.148602	5.517324
1646	VCYT	1.015007	0.057058	2	0.021367	0.238869	4.159773
890	JYNT	0.728127	0.037189	2	0.021367	0.192844	3.664930
640	FRPT	0.921197	0.069974	2	0.021367	0.264525	3.401683
1335	RGEN	0.714409	0.058437	2	0.021367	0.241738	2.866914
1474	SPSC	0.604935	0.046052	2	0.021367	0.214596	2.719375
968	LULU	0.674161	0.057792	2	0.021367	0.240400	2.715447
606	FIVN	0.815347	0.103550	2	0.021367	0.321792	2.467373
81	AMD	1.044455	0.189414	2	0.021367	0.435217	2.350752
484	DXCM	0.794039	0.108452	2	0.021367	0.329321	2.346259

How to get options data for free?

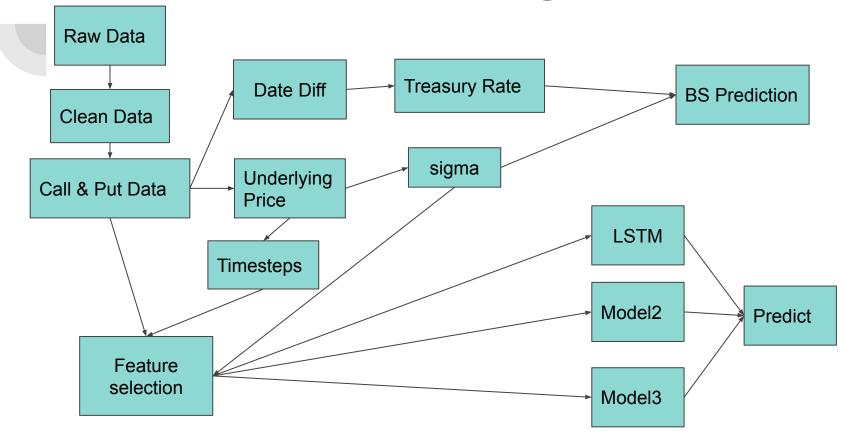
```
from bs4 import BeautifulSoup
  import requests
  import pandas as pd
  from datetime import datetime
  import time
def main():
    print("Hello World!")
    if name == ' main ':
     main()
data url = "https://finance.yahoo.com/quote/SPY/options"
  data html = requests.get(data_url).content
  print(data html)
```

Definitions

df.head()													
Underlying Symbol	UnderlyingPrice	Exchange	Option Symbol	Blank	Туре	Expiration	DataDate	Strike	Last	Bid	Ask	Volume	OpenInterest
ADSK	107.12	w	ADSK180105C00060000	NaN	call	01/05/2018	1/2/2018 04:00:00 PM	60.0	0.0	44.85	49.50	0.0	0.0
P 13		8 70		7		(2)		6.2		ST 42		7	

- * Underlying: The stock, index, or ETF symbol
- * Underlying price: The last traded price at the time of the option quote.
- * Option type: Call or put Expiration The expiration date of the option.
- * Expiration date: The date of the expiration
- * Strike: The strike of the option
- * Bid: The bid price of the option
- * Ask: The ask price of the option
- * Volume: The number of contracts traded
- * Open interest : always a day behind. The OCC changes this number at
- 3:00AM every morning and the number does not change through the day

Data Preprocessing

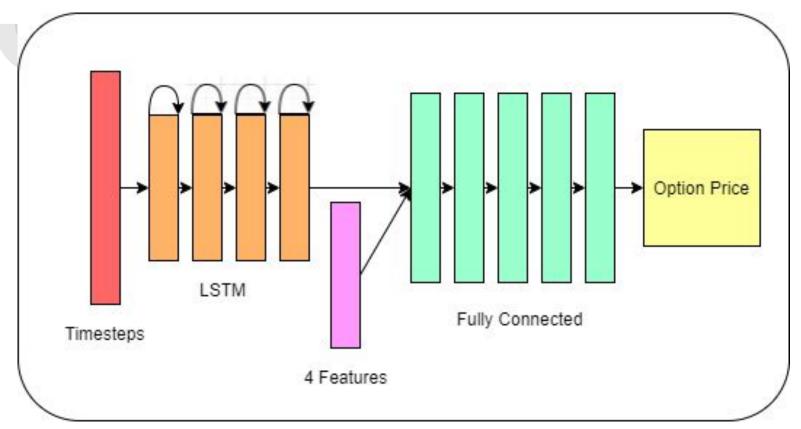


input: [(?, 5)]

LSTM Features Timestep 5

	UnderlyingPrice	Strike	Bid	Ask	date_diff	treasury_rate	1	2	3	4	5
2344	11.91	0.5	11.35	11.45	3	1.33	11.91	12.02	12.1404	11.96	11.82
2345	11.91	1.0	10.90	10.95	3	1.33	11.91	12.02	12.1404	11.96	11.82
2346	11.91	1.5	10.35	10.45	3	1.33	11.91	12.02	12.1404	11.96	11.82
2347	11.91	2.0	9.90	9.95	3	1.33	11.91	12.02	12.1404	11.96	11.82
2348	11.91	2.5	9.35	9.45	3	1.33	11.91	12.02	12.1404	11.96	11.82
	3422	7500	1922	1922	122	3.2	1922	1922	222	925	0.00
702371	148.52	185.0	16.40	19.20	534	2.00	30.45	33.87	33.4800	34.02	33.67
702372	148.52	190.0	15.00	17.80	534	2.00	30.45	33.87	33.4800	34.02	33.67
702373	148.52	195.0	13.90	16.40	534	2.00	30.45	33.87	33.4800	34.02	33.67
702374	148.52	200.0	12.90	15.20	534	2.00	30.45	33.87	33.4800	34.02	33.67
702375	148.52	210.0	10.10	13.20	534	2.00	30.45	33.87	33.4800	34.02	33.67

LSTM



Models-LSTM

call_model.summary()

Model: "functional 1"

Layer (type)	Output	Shape	Param #	Connected to
input_1 (InputLayer)	[(None	, 5, 1)]	0	
sequential (Sequential)	(None,	8)	1952	input_1[0][0]
input_2 (InputLayer)	[(None	, 4)]	0	
concatenate (Concatenate)	(None,	12)	0	sequential[0][0] input_2[0][0]
dense (Dense)	(None,	100)	1300	concatenate[0][0]
batch_normalization (BatchNorma	(None,	100)	400	dense[0][0]
leaky_re_lu (LeakyReLU)	(None,	100)	0	batch_normalization[0][0]
dense_1 (Dense)	(None,	100)	10100	leaky_re_lu[0][0]
batch_normalization_1 (BatchNor	(None,	100)	400	dense_1[0][0]
leaky_re_lu_1 (LeakyReLU)	(None,	100)	0	batch_normalization_1[0][0]
dense_2 (Dense)	(None,	100)	10100	leaky_re_lu_1[0][0]
batch_normalization_2 (BatchNor	(None,	100)	400	dense_2[0][0]
leaky_re_lu_2 (LeakyReLU)	(None,	100)	0	batch_normalization_2[0][0]
dense 3 (Dense)	(None,	1)	101	leaky re lu 2[0][0]

[(?, 5, 1)] input_1: InputLayer output: [(?, 5, 1)] [(?, 4)] (?, 5, 1)sequential: Sequential input_2: InputLayer [(?, 4)] output: (?, 8)[(?, 8), (?, 4)]concatenate: Concatenate (?, 12)(?, 12) dense: Dense (?, 100)(?, 100) input: batch_normalization: BatchNormalization output: (?, 100) input: (?, 100) leaky_re_lu: LeakyReLU output: (?, 100) (?, 100) dense_1: Dense output: input: (?, 100) batch normalization 1: BatchNormalization output: (?, 100) (?, 100) input: leaky_re_lu_1: LeakyReLU output: (?, 100) (?, 100) dense_2: Dense output: (?, 100) input: (?, 100) batch_normalization_2: BatchNormalization output: (?, 100) (?, 100) leaky re lu 2: LeakyReLU (?, 100) (?, 100) dense 3: Dense

Model2/Model3 Features Sigma 5

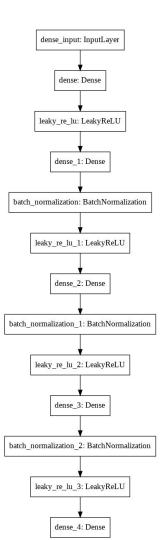
all_d	lf_5.head()						
	UnderlyingPrice	Strike	Bid	Ask	sigma_5	date_diff	treasury_rate
1042	12.28	5.5	6.75	6.85	0.033379	4	1.3
1043	12.28	6.0	6.20	6.35	0.033379	4	1.3
1044	12.28	6.5	5.75	5.85	0.033379	4	1.3
1045	12.28	7.0	5.25	5.35	0.033379	4	1.3
1046	12.28	7.5	4.70	4.85	0.033379	4	1.3

```
for day in [5, 10, 21, 30]:
    df_underlying['sigma_'+str(day)] = df_underlying['UnderlyingPrice'].rolling(day).apply(lambda x: (np.diff(x) / x[:-1]).std())
```

Models- Model2/3

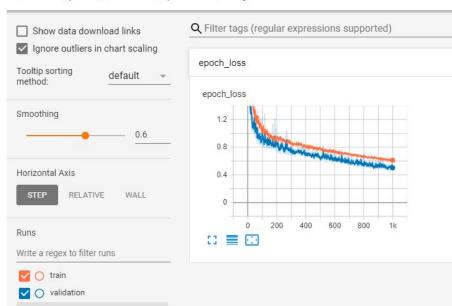
Same architecture as Model2 ,but Model3 predict both ask and bid prices (Two output can be used for options seller).

Layer (type)	Output	Shape	Param #
dense_13 (Dense)	(None,	400)	2400
leaky_re_lu_8 (LeakyReLU)	(None,	400)	0
dense_14 (Dense)	(None,	400)	160400
batch_normalization (BatchNo	(None,	400)	1600
leaky_re_lu_9 (LeakyReLU)	(None,	400)	0
dense_15 (Dense)	(None,	400)	160400
batch_normalization_1 (Batch	(None,	400)	1600
leaky_re_lu_10 (LeakyReLU)	(None,	400)	0
dense_16 (Dense)	(None,	400)	160400
batch_normalization_2 (Batch	(None,	400)	1600
leaky_re_lu_11 (LeakyReLU)	(None,	400)	0
dense_17 (Dense)	(None,	2)	802
Total params: 489,202 Trainable params: 486,802 Non-trainable params: 2,400			



LSTM2_all_put10

n_units = 400,layers = 4,n_batch = 1024,n_epochs = 1000,learning_rate=1e-4



LSTM

Put5

https://tensorboard.dev/experiment/IN9XplUvRbqhaVBvCPIJSg/

Put10

https://tensorboard.dev/experiment/oQljEuYVSvOLPDi17AtRww/#scalars

Put21

https://tensorboard.dev/experiment/h5KIjLjiQPmdk9cvD9JD6A/

Call10

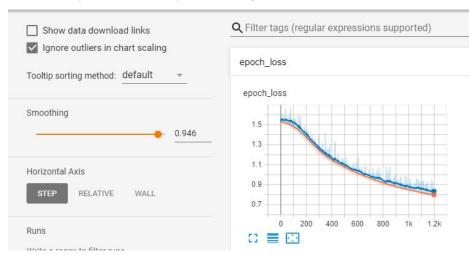
https://tensorboard.dev/experiment/9tG0Hv68TmyaI7NBnbr2Ow/

Call21

https://tensorboard.dev/experiment/ZE4cErmZRXKKvUJRpHbjHQ/#scalars

Model2_all_put

n_units = 400,layers = 4,n_batch = 1024,n_epochs = 2400,learning_rate=1e-5



Model2:

Call 5

https://tensorboard.dev/experiment/Bgx7f38NSgOtJQdNMXI9JA/#scalarscall10

https://tensorboard.dev/experiment/y9MuyU77Tymi0Uz1Yb0NSA/#scalars Call 21

https://tensorboard.dev/experiment/y9MuyU77Tymi0Uz1Yb0NSA/

put5

https://tensorboard.dev/experiment/L208zOrlRziyfgiEm9RjjQ/#scalars Put21

https://tensorboard.dev/experiment/L208zOrlRzivfqiEm9RjjQ/

Model3:

call5

https://tensorboard.dev/experiment/X4dEGe3NQbOBOU5vFzHQuw/Call21

https://tensorboard.dev/experiment/V2WF49BDQIK1spQ21eedfA/

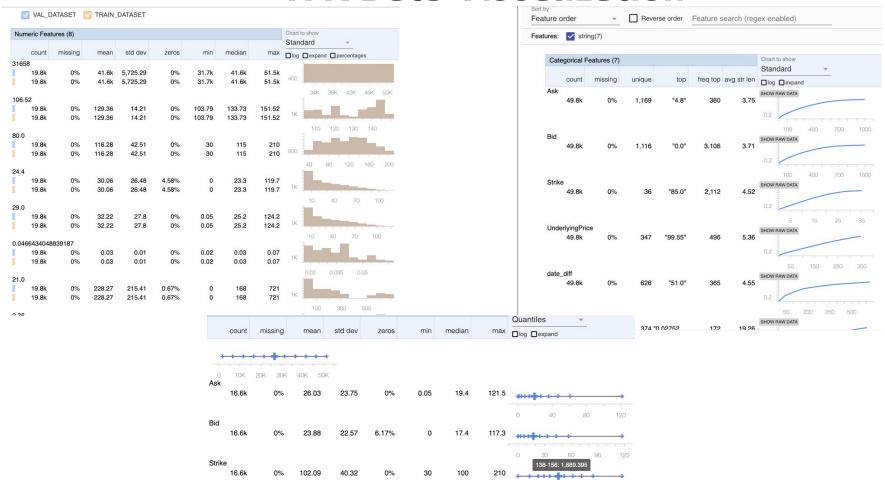
Put5

 $\underline{https://tensorboard.dev/experiment/RY5C4BGKRIKpndYhO4sdqg/}$

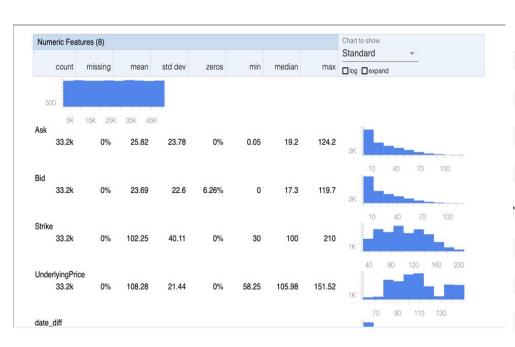
Put21

https://tensorboard.dev/experiment/9zzUE8NcQoySec6DarEo9g/

TFX Data Visualization



Data Schema/Statistics



	Anomaly short description	Anomaly long description
Feature name		
'treasury_rate'	Column dropped	Column is completely missing
'\'29.0\''	New column	New column (column in data but not in schema)
'\'80.0\''	New column	New column (column in data but not in schema)
'\'24.4\''	New column	New column (column in data but not in schema)
'date_diff'	Column dropped	Column is completely missing
'31658'	New column	New column (column in data but not in schema)
'\'106.52\''	New column	New column (column in data but not in schema)
'Bid'	Column dropped	Column is completely missing
0.0466434048839187\"	New column	New column (column in data but not in schema)
'Ask'	Column dropped	Column is completely missing
'UnderlyingPrice'	Column dropped	Column is completely missing
'\'21.0\''	New column	New column (column in data but not in schema)
'sigma_21'	Column dropped	Column is completely missing
'Strike'	Column dropped	Column is completely missing
'\'2.36\''	New column	New column (column in data but not in schema)

TFX Model Serving

```
2020-12-07 00:54:58.210420: I external/org_tensorflow/tensorflow/cc/saved_model/loader.cc:199] Restoring SavedModel bundle.
2020-12-07 00:54:58.254873: I external/org_tensorflow/tensorflow/cc/saved_model/loader.cc:183] Running initialization op on SavedModel bundle a
2020-12-07 00:54:58.262050: I external/org_tensorflow/tensorflow/cc/saved_model/loader.cc:303] SavedModel load for tags { serve }; Status: successfully loaded servables of tensorflow serving/servables/tensorflow/saved_model_warmup_util.cc:59] No warmup data file found at /tmp/1/assets
2020-12-07 00:54:58.263216: I tensorflow_serving/core/loader_harness.cc:87] Successfully loaded servable version {name: first_model version: 1}
2020-12-07 00:54:58.264399: I tensorflow_serving/model_servers/server.cc:367] Running gRPC ModelServer at 0.0.0.0:8500 ...
[warn] getaddrinfo: address family for nodename not supported
2020-12-07 00:54:58.265214: I tensorflow_serving/model_servers/server.cc:387] Exporting HTTP/REST API at:localhost:8501 ...
```

Predictions

'{"signature name": "serving default", "instances": [[141.96, 190.0, 0.040351047419522064, 225.0, 2.22], [82.85, 70.0, 0.02629 471439129068, 37.0, 1.64], [97.4, 155.0, 0.03228772966546013, 571.0, 2.54], [110.11, 135.0, 0.06315234590095266, 42.0, 2.41], [109.1, 95.0, 0.016403341833589692, 140.0, 2.52], [137.46, 115.0, 0.026066182636784167, 4.0, 2.41], [79.65, 60.0, 0.0264284369 79249875, 228.0, 1.86], [105.2, 130.0, 0.02488153995771937, 178.0, 2.19], [79.35, 75.0, 0.02112959870762864, 318.0, 2.06], [9 7.4, 45.0, 0.03228772966546013, 116.0, 1.93], [111.96, 70.0, 0.04309717861046268, 17.0, 2.14], [106.3, 60.0, 0.036272658671604 59, 15.0, 1.98], [96.6, 135.0, 0.06593577053311157, 63.0, 2.36], [90.65, 145.0, 0.05369590696567412, 744.0, 2.5], [108.15, 95. 0, 0.01953874168830817, 130.0, 2.46], [101.34, 120.0, 0.05211642321528565, 93.0, 2.45], [89.4, 40.0, 0.02907330589461383, 93.0]0, 1.81], [141.96, 55.0, 0.040351047419522064, 225.0, 2.22], [125.75, 65.0, 0.0567472996325723, 696.0, 2.5], [122.65, 170.0, 0.04381723541187575, 191.0, 2.46], [112.0, 75.0, 0.044291377964058416, 480.0, 2.6], [95.0, 60.0, 0.039871988179493965, 535.0, 2.44], [142.03, 190.0, 0.02304807360549109, 256.0, 2.46], [119.3, 60.0, 0.041100395508881706, 108.0, 2.42], [107.6, 145.0, 0.0 2658037020883468, 576.0, 2.56], [82.05, 40.0, 0.03607495935958746, 248.0, 2.09], [106.3, 75.0, 0.03627265867160459, 15.0, 1.9 8], [124.48, 95.0, 0.035416602172033324, 276.0, 2.45], [83.425, 30.0, 0.03688202209474797, 254.0, 2.05], [146.01, 110.0, 0.040 34332393473633, 11.0, 2.3], [110.85, 90.0, 0.02266813029334776, 86.0, 2.01], [83.18, 70.0, 0.05119926052739802, 4.0, 2.36], [1 08.89, 85.0, 0.033616104014165364, 59.0, 2.46], [139.47, 100.0, 0.02178936629123049, 549.0, 1.87], [77.4, 35.0, 0.0229767343221936, 291.0, 2.08], [144.62, 155.0, 0.02990831354276945, 108.0, 2.2], [112.45, 65.0, 0.046253993306802715, 485.0, 2.58], [148. 12, 105.0, 0.03955871329297976, 28.0, 2.161, [140.94, 80.0, 0.036935548741476684, 590.0, 1.83], [86.45, 110.0, 0.0358836905231

MSE Results

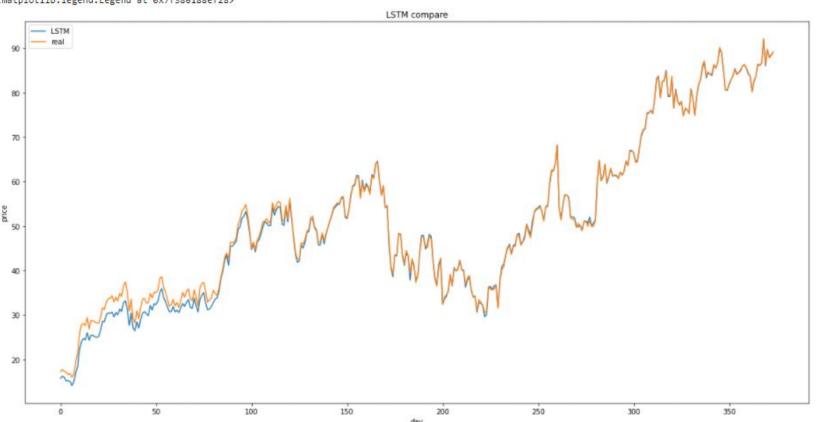
model	learning rate	Batch Size	epoch	sigma	MSE
LSTM call	0.0001	1024	1000	21	0.58
LSTM put	0.0001	1024	1000	10	0.6089
Model2 call	1.00E-05	1024	4000	21	0.8284
Model2 put	1.00E-05	1024	3000	5	0.821
Model3 call	1.00E-05	1024	2000	5	0.4618
Model3 put	1.00E-05	1024	4000	21	0.8204

model	sigma5	sigma10	sigma21
LSTM call	0.8	0.6206	0.58
LSTM put	0.69	0.6089	0.6206
Model2 call	0.833	0.8442	0.8284
Model2 put	0.821	0.9118	0.832
Model3 call	0.4618	0.8623	0.7406
Model3 put	0.8602	1.0232	0.8204
BS Model call	Nan	Nan	23.2528235
Bs Model put	Nan	Nan	23.7103045

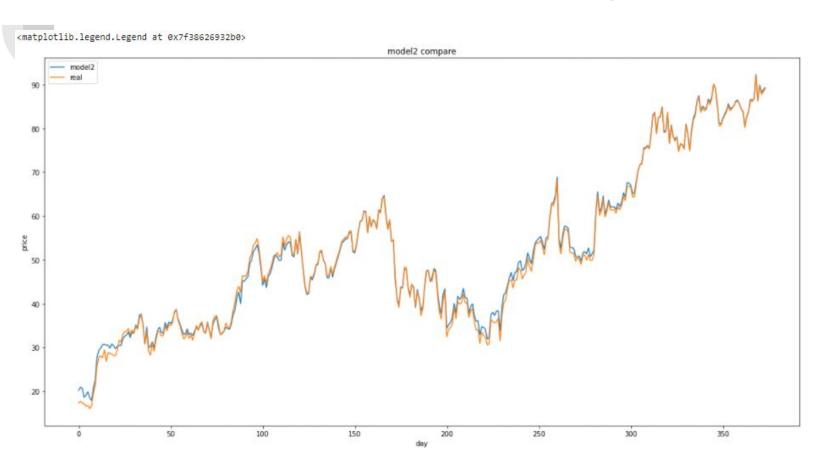
model	MSE
LSTM call	0.58
LSTM put	0.6089
Model2 call	0.8284
Model2 put	0.832
Model3 call	0.4618
Model3 put	0.8204
BS Model call	23.2528235
Bs Model put	23.7103045

LSTM Mean Price Prediction

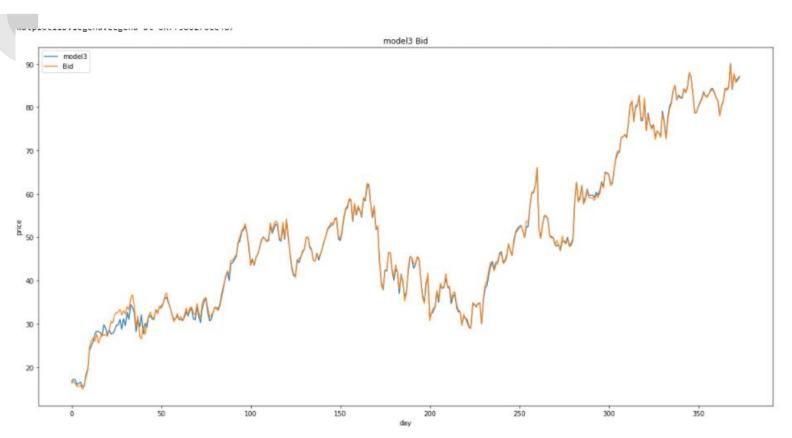
<matplotlib.legend.Legend at 0x7f386188ef28>



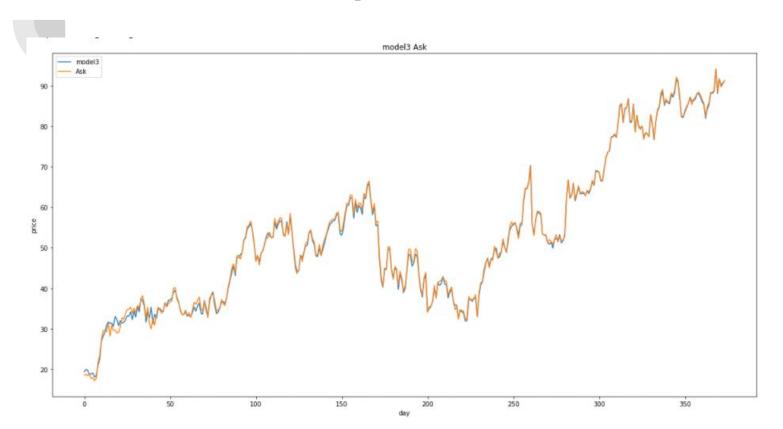
Model2 Mean Price Prediction

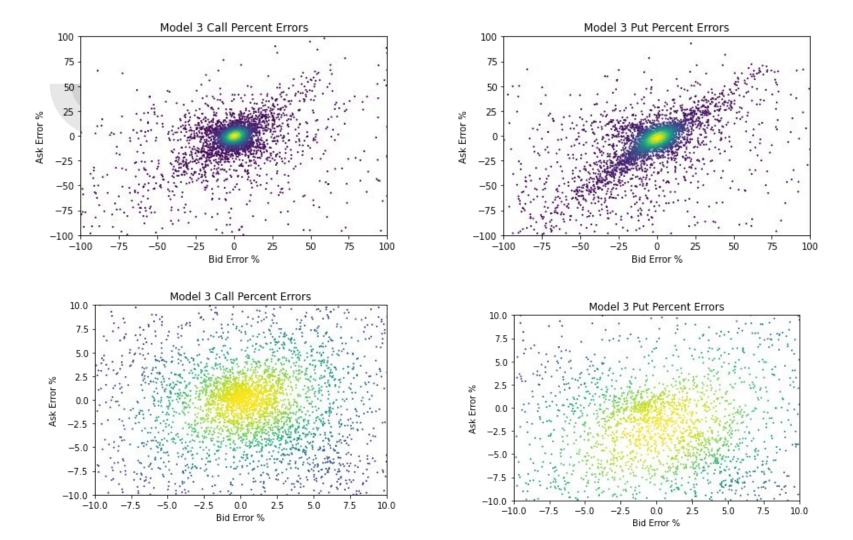


Model 3 Bid price Prediction



Model 3 on Ask price Prediction







https://github.com/zjzsu2000/CMPE297_AdvanceDL_Project

Future Work

- Stocks filter Using the stocks with better Sharpe Rate
- Option filter
 - Not using the options deep out of the money(The Deep-OTM options are more volatile)
 - Not using the options expired in 7 days (The options closer to the expiry date are more volatile)
- More models, More data, More training, More metric
- Apply to the real trading
- Fine-tuning with more hyperparameters