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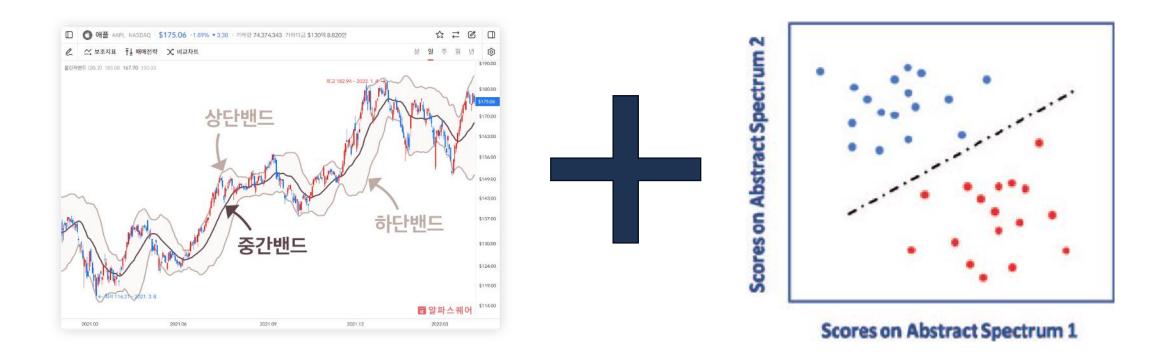
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- 2 프로젝트 과정
- 3 프로젝트 결과

Part 1

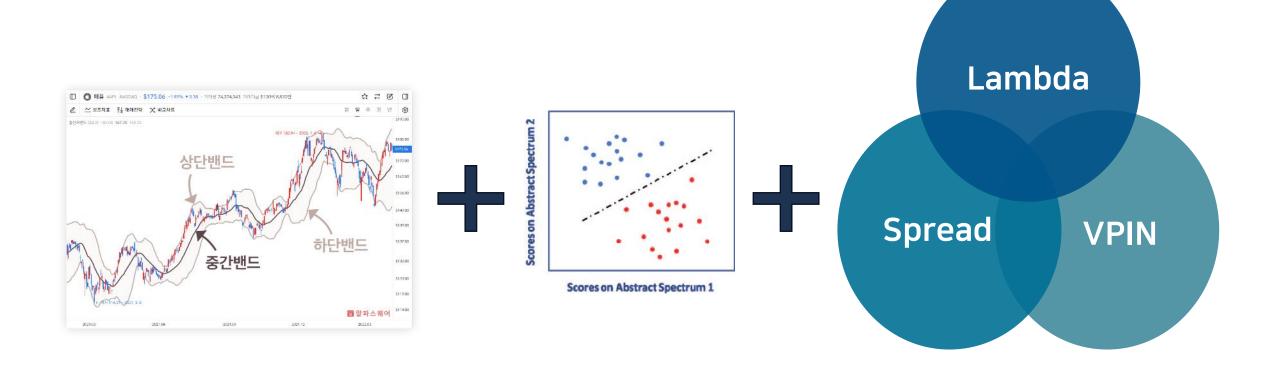
프로젝트동기



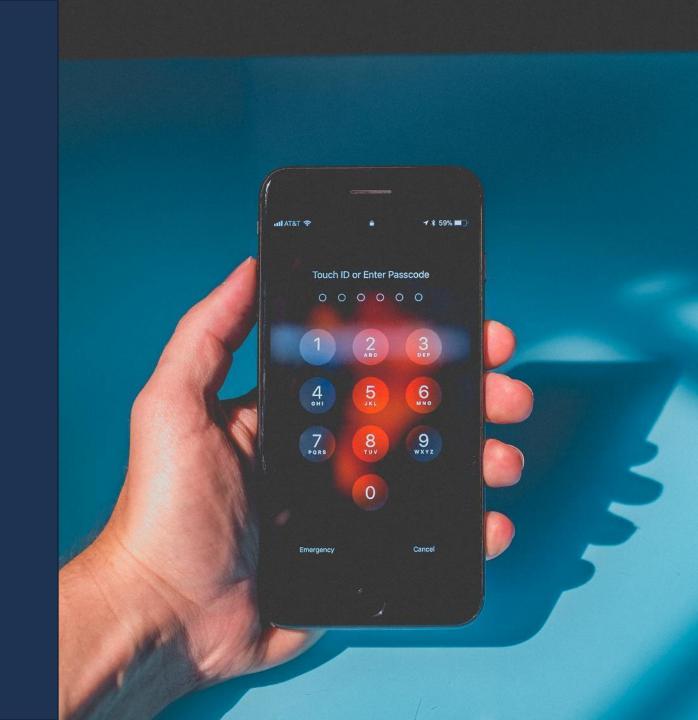
프로젝트 동기



프로젝트 동기



Part 2 프로젝트과정



STEP 1		STEP 2		STEP 3		STEP 4
데이터준비및전처리	>>	볼린저밴드를이용한 거래신호생성모델구현 (primary model)	>>	거래신호를 meta labeling 한결과물을 분류하는 모델 구현 (secondary model)	>>	분류모델을이용한 Backtesting 수행

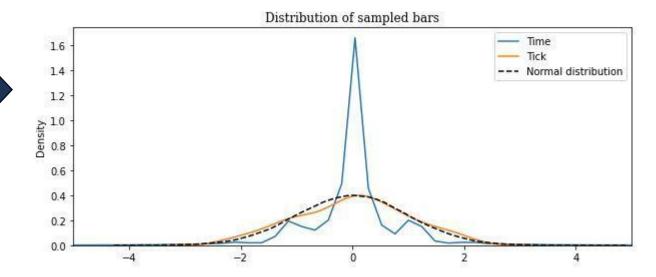
STEP 1

데이터준비및전처리

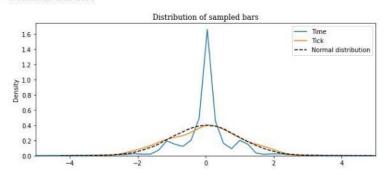
bar = StandardBarFeatures(file_path_or_df = df)
dollar = bar.dollar_bar(threshold =
int(np.mean(df.close*df.volume))*1000)

Reading data in batches: Batch number: 0 Returning bars

0.035834167152643204



0.035834167152643204



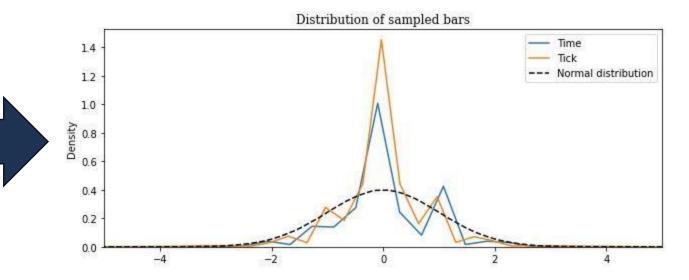
	date	time	open	high	low	close	volume
0	20211229	1106	14800	14800	14800	14800	21
1	20211229	1403	14750	14750	14750	14750	3
2	20211229	1404	14700	14750	14700	14750	5
3	20211229	1405	14700	14750	14700	14700	1291
4	20211229	1406	14750	14750	14750	14750	4
	***	9.0		1900	***	440	
136995	20231108	924	9810	9820	9810	9820	123
136996	20231108	923	9820	9820	9820	9820	8
136997	20231108	922	9830	9830	9830	9830	80
136998	20231108	941	9810	9810	9810	9810	241
136999	20231108	1029	9840	9840	9840	9840	10



	date_time	tick_num	open	high	low	close	volume	cum_buy_volume	cum_ticks	cum_dollar_value
37	20220222	11937	1142.0	1530.0	901.0	1126.0	4172350	893400	327	4.954299e+09
38	20220223	12263	1124.0	1530.0	901.0	1143.0	4162650	956100	326	4.957347e+09
39	20220224	12600	1142.0	1530.0	901.0	1028.0	4196650	958450	337	4.953505e+09
40	20220228	12933	1047.0	1530.0	901.0	1325.0	4124650	1054600	333	4.950348e+09
41	20220228	13262	1347.0	1530.0	901.0	1107.0	4140100	831300	329	4.954124e+09
	196	144	, en	***	***	(***)		644	(149)	***
339	20231027	134918	1102.0	1530.0	901.0	1452.0	4071890	776190	430	4.953774e+09
340	20231030	135349	1451.0	1530.0	901.0	1149.0	4138830	1939670	431	4.951925e+09
341	20231101	135776	1151.0	1530.0	901.0	1330.0	4135500	900350	427	4.949941e+09
342	20231103	136217	1329.0	1530.0	901.0	1300.0	4235860	1894040	441	4.956160e+09
343	20231106	136654	1303.0	1530.0	901.0	1049.0	4243370	3574070	437	4.952413e+09

307 rows × 22 columns

tick=bar.volume_bar(threshold = 50000) # np.mean(df.volume)=29087



	date_time	tick_num	open	high	low	close	volume	cum_buy_volume	cum_ticks	cum_dollar_value
0	2017-04-24 09:03:00	2	41200.0	41200.0	41200.0	41200.0	82420	0	2	3.395704e+09
1	2017-04-24 09:05:00	4	41200.0	41200.0	41200.0	41200.0	82420	0	2	3.395704e+09
2	2017-04-24 09:07:00	6	41140.0	41140.0	41140.0	41140.0	82340	0	2	3.387468e+09
3	2017-04-24 09:09:00	8	41080.0	41080.0	41080.0	41080.0	82200	0	2	3.376776e+09
4	2017-04-24 09:11:00	10	41100.0	41120.0	41100.0	41120.0	82240	82240	2	3.380886e+09
		100	225		239			***	***	1999
118122	2019-05-15 15:13:00	188717	42700.0	42700.0	42650.0	42650.0	85450	0	2	3.646580e+09
118123	2019-05-15 15:15:00	188719	42650.0	42700.0	42650.0	42700.0	85400	42700	2	3.644445e+09
118124	2019-05-15 15:17:00	188721	42600.0	42700.0	42600.0	42700.0	85400	42700	2	3.642310e+09
118125	2019-05-15 15:19:00	188723	42650.0	42700.0	42650.0	42700.0	85400	42700	2	3.644445e+09
118126	2019-05-15 15:30:00	188725	42650.0	42650.0	42550.0	42550.0	85250	0	2	3.631658e+09

118127 rows x 10 columns

```
tick['mfi']=money_flow_index(high = tick.high,
                           low = tick.low,
                           close = tick.close,
                           volume = tick.volume,
                           window = 20
tick['obv']=on_balance_volume(close=tick.close,
                           volume=tick.volume,
tick['rsi']=rsi(close=tick.close, window=20)
tick['vwap']=volume_weighted_average_price(high=tick.high,
                                        low=tick.low,
                                        close=tick.close,
                                        volume=tick.volume,
                                        window=20)
tick['tsi']=tsi(close=tick.close,
             window_fast=13,
             window_slow=25)
```

```
from FinancialMachineLearning.features.microstructure import *

corwin_schultz = CorwinSchultz(tick['high'], tick['low'])
spread = corwin_schultz.corwin_schultz_estimator(window = 20)
```

```
from FinancialMachineLearning.barsampling.bar_feature import BarFeature
def buy volume(df):
    tick_signs = tick_rule(df['price'])
    return (df['volume'] * (tick signs > 0)).sum()
def sell volume(df):
    tick_signs = tick_rule(df['price'])
    return (df['volume'] * (tick_signs < 0)).sum()
buy_volume_feature = BarFeature(name='buy_volume', function=buy_volume)
sell_volume_feature = BarFeature(name='sell_volume', function=sell_volume)
bars = vpin_volume_bars('./tick_bar.csv', additional_features = [buy_volume_feature, sell_volume_feature])
vol thres = 10000
vpin_series = vpin(bars['volume'], bars['buy_volume'], window = 5)
```

STEP 2

볼린저밴드를이용한 거래신호생성모델구현 (primary model)

```
bollinger_result=BollingerBands(close = tick["close"], window = 20, window_dev = 1.2)
tick['boll_hband']=np.array(bollinger_result.bollinger_hband_indicator()).astype(int)
tick['boll lband']=np.array(bollinger result.bollinger lband indicator())
label=[]
for indexs in tick index:
   if tick.loc[indexs, 'boll_hband']==1:
       label.append(1)
   elif tick.loc[indexs, 'boll_lband']==1:
       label.append(-1)
   else:
       label.append(0)
```

bollinger_result=BollingerBands(close = tick["close"], window = 20, window_dev = 2)

	precision	recall	f1-score	support
0	0.94	0.99	0.96	21066
1	0.82	0.51	0.63	2552
accuracy			0.93	23618
macro avg	0.88	0.75	0.80	23618
weighted avg	0.93	0.93	0.93	23618

bollinger_result=BollingerBands(close = tick["close"], window = 20, window_dev = 0.5)

	precision	recall	f1-score	support
0 1	0.54 0.79	0.30 0.91	0.39 0.85	6038 17566
accuracy macro avg weighted avg	0.67 0.73	0.61 0.76	0.76 0.62 0.73	23604 23604 23604

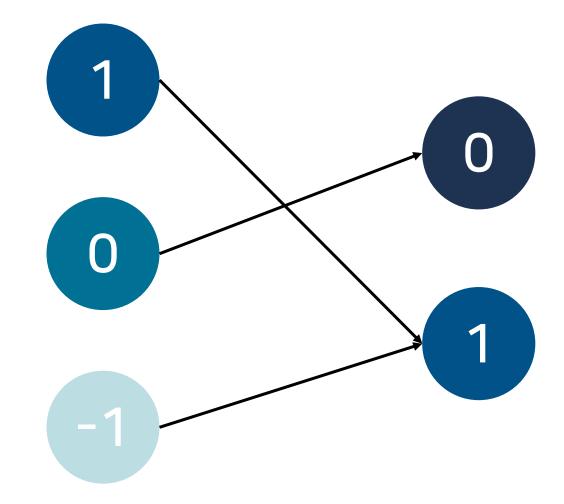
bollinger_result=BollingerBands(close = tick["close"], window = 20, window_dev = 1.2)

	precision	recall	f1-score	support
0	0.84	0.91	0.87	14739
1	0.82	0.71	0.76	8879
accuracy			0.83	23618
macro avg	0.83	0.81	0.82	23618
weighted avg	0.83	0.83	0.83	23618

STEP 3

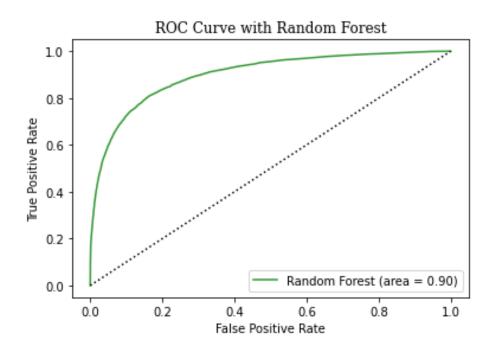
거래신호를 meta labeling 한결과물을 분류하는 모델 구현 (secondary model)

```
# meta labling
meta=[]
for data in tick.label:
    if data==1:
        meta.append(1)
    elif data==-1:
        meta.append(1)
    else:
        meta.append(0)
tick['meta']=meta
```



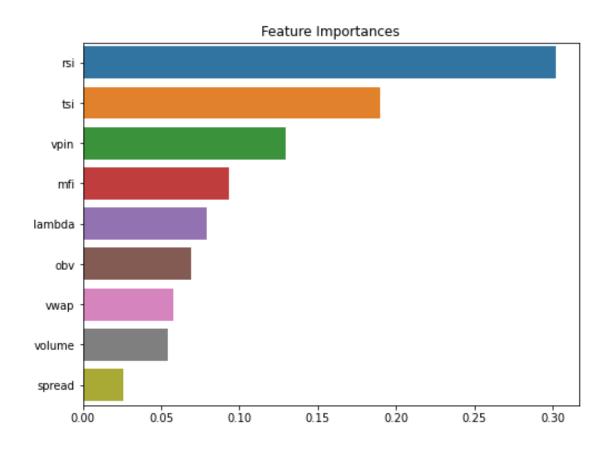
```
from sklearn.model_selection import train_test_split
from sklearn, ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
from sklearn metrics import roc auc score, classification report, roc curve
from sklearn metrics import auc
tick=tick.dropna()
X=tick[['spread', 'lambda', 'mfi', 'obv', 'rsi', 'vwap', 'tsi', 'volume', 'vpin']]
y=tick['meta']
X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.2, shuffle=False)
tree = RandomForestClassifier(
        criterion='entropy', class weight='balanced subsample', min weight fraction leaf=0.0, n estimators=500,
       max features=1.0, oob score=True, n jobs=-1)
tree.fit(X train, y train)
y pred tree = tree.predict(X test)
```





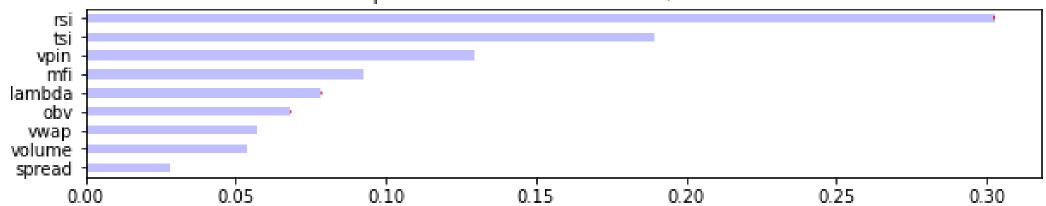
	precision	recall	f1-score	support	
0 1	0.84 0.82	0.91 0.71	0.87 0.76	14739 8879	
accuracy macro avg weighted avg	0.83 0.83	0.81 0.83	0.83 0.82 0.83	23618 23618 23618	

ser = pd.Series(tree.feature_importances_, index=feature_name)

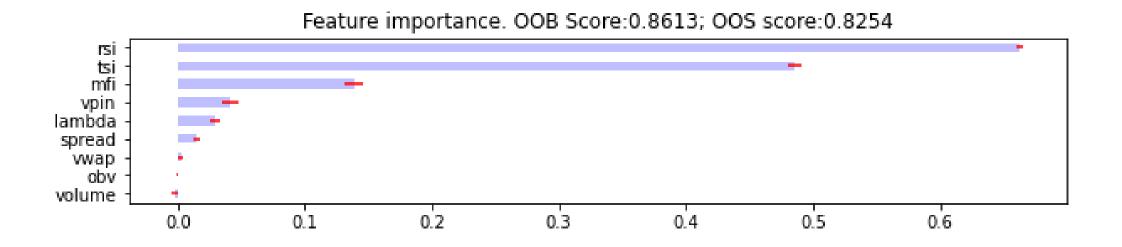


test_data_func(X, tick, methods='MDI')





test_data_func(X, tick, methods='MDA')



```
start=datetime.datetime(2008,1,1)
end=datetime.datetime(2012,12,31)
data=pdr.DataReader("005930", "naver", start, end).astype("float")
```

```
def bollinger hband(df):
   bollinger_result=BollingerBands(close = pd.Series(df.Close), window = 20, window_dev = 1)
   boll_hband=np.array(bollinger_result.bollinger_hband_indicator()).astype(int)
   label=[]
   for data in boll_hband:
       if data==1:
           label.append(1)
       else.
            label.append(0)
   return label
def bollinger lband(df):
   bollinger_result=BollingerBands(close = pd.Series(df.Close), window = 20, window_dev = 1)
   boll_lband=np.array(bollinger_result.bollinger_lband_indicator()).astype(int)
   label=[]
   for data in boll lband:
       if data==1:
           label.append(-1)
       else:
           label.append(0)
   return label
def spread(high, low):
   corwin_schultz = CorwinSchultz(pd.Series(high), pd.Series(low))
   spread = corwin_schultz.corwin_schultz_estimator(window = 20)
   spread=spread.fillna(0)
   return spread
def lambda_data(close, volume):
   lambda_feature = BarbasedLambda(close = pd.Series(close),
                                   volume = pd.Series(volume),
                                   dollar_volume = tick.close * tick.volume)
   kyle lambda = lambda feature.kyle()
   kyle_lambda=kyle_lambda.fillna(0)
   return kyle lambda
```

```
def model predict(spread, kyle lambda, mfi, oby, rsi return, ywap, tsi return, yolume):
   data=pd.DataFrame({'spread': spread, 'kyle':kyle_lambda, 'mfi':mfi, 'obv': obv, 'rsi':rsi_return, 'wwap':vwap, 'tsi':tsi_return, 'volum
    data=data.fillna(0)
   y_pred_tree = tree.predict(data)
   return y_pred_tree
def model_prob(spread, kyle_lambda, mfi, obv, rsi_return, vwap, tsi_return, volume):
    data=pd.DataFrame({'spread': spread, 'kyle':kyle_lambda, 'mfi':mfi, 'obv': obv, 'rsi':rsi_return, 'vwap':vwap, 'tsi':tsi_return, 'volum
   data=data.fillna(0)
   y_pred_prob = tree.predict_proba(data)[:,1]
   return y_pred_prob
   return money_flow_index(high = pd.Series(data.High),
                           low = pd.Series(data.Low).
                          close = nd.Series(data,Close).
                           volume = pd.Series(data.Volume),
def_obv(data):
   return on_balance_volume(close=pd.Series(data.Close),
                          volume=pd.Series(data.Volume))
def rsi_return(data):
   return rsi(close=pd.Series(data.Close), window=20)
   return volume_weighted_average_price(high=pd.Series(data.High),
                                       low=pd.Series(data.Low),
                                       close=pd.Series(data.Close),
                                       volume=pd.Series(data.Volume),
                                       window=20)
def tsi_return(data):
    return tsi(close=pd.Series(data.Close),
            window fast=13,
            window_slow=25)
```

```
class my_strategy(Strategy):
    def init(self):
        self.bollinger_h = self.I(bollinger_hband, self.data)
        self.bollinger_l = self.I(bollinger_lband, self.data)
        self.bollinger=self.bollinger h+self.bollinger l
        self.spread=self.I(spread, self.data.High, self.data.Low)
        self.lambda lst=self.I(lambda data, self.data.Close, self.data.Volume)
        self.mfi=self.I(mfi, self.data)
        self.obv=self.I(obv, self.data)
        self.rsi=self.I(rsi_return, self.data)
        self.vwap=self.I(vwap, self.data)
        self.tsi=self.I(tsi_return, self.data)
        self.volume=self.data.Volume
        self.pred=self.I(model_predict, self.spread, self.lambda_lst, self.mfi, self.obv, self.rsi
        self.prob=self.I(model_prob, self.spread, self.lambda_lst, self.mfi, self.obv, self.rsi, s
        self.result=self.bollinger*self.pred
    def next(self):
       if self.result==1:
            self.buy( size=(self.prob[-1]) )
        elif self.result==-1:
            self.position.close( portion=(self.prob[-1]) )
bt = Backtest(data, my_strategy, cash=10000000, commission=.002)
stats = bt.run()
bt.plot()
```



