

Applying a Random Forest

MGMT 638: Data-Driven Investments: Equity

Kerry Back, Rice University



Outline

- Create current features:
 - Get data from SQL library.
 - We only want most recent data, but go back a couple of years to compute momentum, growth rates, etc.
 - Follow same procedure as in 5a-fundamentals.ipynb, but do not shift momentum, volatility, etc. forward.
 - And do not keep return (return for prior week is not useful)
- Apply saved random forest model to current data to form future predictions.
- Use predictions to identify best and worst stocks today (maybe sector neutral).



Create connection



```
In [18]: import pandas as pd

from sqlalchemy import create_engine
import pymssql
server = 'fs.rice.edu'
database = 'stocks'
username = 'stocks'
password = '6LAZH1'
string = "mssql+pymssql://"+ username + ":" + password + "@" + server + "/"
conn = create_engine(string).connect()
```

```
Exception during reset or similar
Traceback (most recent call last):
  File "c:\Users\kerry\AppData\Local\Programs\Python\Python310\lib\site-packages\sqlalchemy\pool\base.py", line 753, in _finalize_fairy
    fairy._reset(pool)
  File "c:\Users\kerry\AppData\Local\Programs\Python\Python310\lib\site-packages\sqlalchemy\pool\base.py", line 1004, in _reset
    pool._dialect.do_rollback(self)
  File "c:\Users\kerry\AppData\Local\Programs\Python\Python310\lib\site-packages\sqlalchemy\dialects\mssql\base.py", line 2792, in do_rollback
    super(MSDialect, self).do_rollback(dbapi_connection)
  File "c:\Users\kerry\AppData\Local\Programs\Python\Python310\lib\site-packages\sqlalchemy\engine\default.py", line 683, in do_rollback
    dbapi_connection.rollback()
  File "src\pymssql\_pymssql.pyx", line 316, in pymssql._pymssql.Connection.rollback
  File "src\pymssql\_pymssql.pyx", line 300, in pymssql._pymssql.Connection.rollback
```



Calculate financial ratios and growth rates

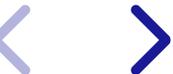
Data from SF1



```
In [19]: sf1 = pd.read_sql(
        """
        select ticker, datekey, lastupdated, netinc, ncfo, equity, assets
        from sf1
        where dimension='ARQ' and datekey>='2021-01-01' and equity>0 and assets>0
        order by ticker, datekey
        """,
        conn,
        parse_dates=["datekey"]
    )
sf1 = sf1.groupby(["ticker", "datekey", "lastupdated"]).last()
sf1 = sf1.droplevel("lastupdated")
sf1 = sf1.reset_index()
```



```
In [ ]: for col in ["netinc", "ncfo"]:
        sf1[col] = sf1.groupby("ticker", group_keys=False)[col].apply(
            lambda x: x.rolling(4).sum()
        )
    for col in ["equity", "assets"]:
        sf1[col] = sf1.groupby("ticker", group_keys=False)[col].apply(
            lambda x: x.rolling(4).mean()
        )
    sf1["roe"] = sf1.netinc / sf1.equity
    sf1["accruals"] = (sf1.netinc - sf1.ncfo) / sf1.equity
    sf1["agr"] = sf1.groupby("ticker", group_keys=False)["assets"].pct_change()
    sf1 = sf1[["ticker", "datekey", "roe", "accruals", "agr"]].dropna()
```



Returns, volume, momentum, volatility

Data from sep_weekly



```
In [ ]: sep_weekly = pd.read_sql(
        """
        select ticker, date, volume, closeadj, closeunadj, lastupdated
        from sep_weekly
        where date>='2022-01-01'
        order by ticker, date, lastupdated
        """,
        conn,
        parse_dates=["date"]
    )
sep_weekly = sep_weekly.groupby(["ticker", "date", "lastupdated"]).last()
sep_weekly = sep_weekly.droplevel("lastupdated")
```



```
In [ ]: sep_weekly["ret"] = sep_weekly.groupby("ticker", group_keys=False).closeadj.p
sep_weekly["annual"] = sep_weekly.groupby("ticker", group_keys=False).closeadj
sep_weekly["monthly"] = sep_weekly.groupby("ticker", group_keys=False).closeadj
sep_weekly["mom"] = sep_weekly.groupby("ticker", group_keys=False).apply(
    lambda d: (1+d.annual)/(1+d.monthly) - 1
)
sep_weekly["volatility"] = sep_weekly.groupby("ticker", group_keys=False).ret
    lambda x: x.rolling(26).std()
)
sep_weekly = sep_weekly[["mom", "volume", "volatility", "closeunadj"]]
sep_weekly = sep_weekly.reset_index()
```



Get marketcap and pb

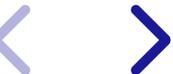
Data from weekly



```
In [ ]: weekly = pd.read_sql(
        """
        select ticker, date, marketcap, pb, lastupdated
        from weekly
        where date>='2022-01-01' and marketcap>0 and pb>0
        order by ticker, date, lastupdated
        """,
        conn,
        parse_dates=["date"]
    )
weekly = weekly.groupby(["ticker", "date", "lastupdated"]).last()
weekly = weekly.droplevel("lastupdated")
weekly = weekly.reset_index()
```



Merge



```
In [ ]: df = weekly.merge(sep_weekly, on=["ticker", "date"], how="inner")
df["year"] = df.date.apply(lambda x: x.isocalendar()[0])
df["week"] = df.date.apply(lambda x: x.isocalendar()[1])
sf1["year"] = sf1.datekey.apply(lambda x: x.isocalendar()[0])
sf1["week"] = sf1.datekey.apply(lambda x: x.isocalendar()[1])
df = df.merge(sf1, on=["ticker", "year", "week"], how="left")
df = df.drop(columns=["year", "week", "datekey"])
```



Fill ratios and growth rates forward



```
In [ ]: for col in ["roe", "accruals", "agr"]:  
        df[col] = df.groupby("ticker", group_keys=False)[col].apply(  
            lambda x: x.ffill()  
        )
```



Add sector data



```
In [ ]: tickers = pd.read_sql(
        """
        select ticker, sector from tickers
        """,
        conn
    )
df = df.merge(tickers, on="ticker")
```



Filter to today's data



```
In [ ]: df = df[df.date==df.date.max()].copy()
```



Filter to small caps and exclude penny stocks



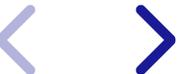
```
In [ ]: df = df[df.closeunadj>5]
df = df.dropna()
df["rnk"] = df.marketcap.rank(
    ascending=False,
    method="first"
)
df = df[(df.rnk>1000) & (df.rnk<=3000)]
df = df.drop(columns=["closeunadj", "rnk"])
```



Define features



```
In [ ]: features = [  
    "marketcap",  
    "pb",  
    "mom",  
    "volume",  
    "volatility",  
    "roe",  
    "accruals"  
]
```



Make predictions



```
In [ ]: # change this to "./" if forest.joblib is in your working directory
path_to_file = "../.."

from joblib import load
forest = load(path_to_file + "forest.joblib")
df["predict"] = forest.predict(X=df[features])
```



Find best and worst stocks



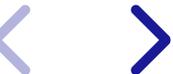
```
In [ ]: df["rnk_long"] = df.predict.rank(  
        ascending=False,  
        method="first"  
    )  
df["rnk_short"] = df.predict.rank(  
    ascending=True,  
    method="first"  
)  
longs = df[df.rnk_long<=44]  
shorts = df[df.rnk_short<=44]
```



Sector-neutral version



```
In [ ]: df["rnk_long"] = df.groupby("sector", group_keys=False).predict.rank(
        ascending=False,
        method="first"
    )
df["rnk_short"] = df.groupby("sector", group_keys=False).predict.rank(
        ascending=True,
        method="first"
    )
longs_neutral = df[df.rnk_long<=4]
shorts_neutral = df[df.rnk_short<=4]
```



Save results



```
In [ ]: with pd.ExcelWriter("portfolios 2023-11-08.xlsx") as writer:
        longs.to_excel(writer, "long", index=False)
        shorts.to_excel(writer, "short", index=False)
        longs_neutral.to_excel(writer, "long neutral", index=False)
        shorts_neutral.to_excel(writer, "short neutral", index=False)
        df.to_excel(writer, "today", index=False)
```

