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 [1]: 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()
```





Calculate financial ratios and growth rates

Data from SF1











Returns, volume, momentum, volatility

Data from sep_weekly











Get marketcap and pb

Data from weekly





```
In [6]: 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 [7]:
    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









Add sector data









Filter to today's data





```
In [10]: df = df[df.date==df.date.max()].copy()
```





Filter to small caps and exclude penny stocks





```
In [11]:
    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"])</pre>
```





Define features









Make predictions





```
In [14]: # 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







Sector-neutral version







Save results





```
in [17]: 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)
```