

Factors

- Value
- Momentum
- Quality
 - Profitability
 - Low accruals
 - Low asset growth
 - Low default probability
- Volatility (low vol and/or low idiosyncratic vol)
- Liquidity (high volume)





Data

- closeadj, closeunadj, volume from sep_weekly
- marketcap, pb from weekly
- netinc, equity, assets, ncfo from sf1 where dimension="ARQ"



Financial Statement Variables

- Use trailing 4 quarters:
 - netinc, ncfo = sum of prior 4 quarters
 - equity, assets = average of prior 4 quarters
- Variables:
 - roe = netinc / equity
 - accruals = (netinc ncfo) / equity
 - agr = % change in assets





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





```
In [2]: sf1 = pd.read_sql(
            select ticker, datekey, lastupdated, netinc, ncfo, equity, assets
            from sf1
            where dimension='ARQ' and datekey>='2009-01-01' 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 [3]:
    for col in ["netinc", "ncfo"]:
        sf1[col] = sf1.groupby("ticker", group_keys=False)[col].apply(lambda x: x
    for col in ["equity", "assets"]:
        sf1[col] = sf1.groupby("ticker", group_keys=False)[col].apply(lambda x: x

    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, momentum, volatility, closeunadj

Data from sep_weekly





```
In [4]: sep_weekly = pd.read_sql(
    """
    select ticker, date, closeadj, closeunadj, lastupdated
    from sep_weekly
    where date>='2010-01-01'
    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 [5]:
    sep_weekly["ret"] = sep_weekly.groupby("ticker", group_keys=False).closeadj.pg
    sep_weekly["annual"] = sep_weekly.groupby("ticker", group_keys=False).closeadg
    sep_weekly["monthly"] = sep_weekly.groupby("ticker", group_keys=False).closeadg
    sep_weekly["mom"] = sep_weekly.groupby("ticker", group_keys=False).apply(
        lambda d: (1+d.annual)/(1+d.monthly) - 1
    )
    sep_weekly["vol"] = sep_weekly.groupby("ticker", group_keys=False).ret.apply()
    sep_weekly = sep_weekly[["ret", "mom", "vol", "closeunadj"]].dropna().reset_index.pdf
```





Get marketcap and pb

Data from weekly









Merge data





```
In [7]:
    df = weekly.merge(sep_weekly, on=["ticker", "date"], how="inner")
    for col in ["pb", "mom", "vol", "marketcap", "closeunadj"]:
        df[col] = df.groupby("ticker", group_keys=False)[col].shift()
        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"])
```



Filter to small caps and exclude penny stocks





```
In [8]: df = df[df.closeunadj>5]
    df["rnk"] = df.groupby("date").marketcap.rank(
        ascending=False,
        method="first"
    )
    df = df[(df.rnk>1000) & (df.rnk<=3000)]
    df = df.reset_index().set_index(["ticker", "date"])</pre>
```





Out[10]:

		index	pb	ret	mom	vol	roe	accruals	
ticker	date								
AACH	2017- 05-05	91	1.0	-0.008357	-0.577655	0.127541	-0.010791	-0.012126	(
	2017- 08-04	92	1.0	-0.062780	-0.687003	0.056854	-0.027568	-0.040935	C
	2018- 11-09	97	1.0	-0.188073	-0.095122	0.066100	-0.220314	-0.125066	C
AAIC	2015- 05-08	143	0.8	-0.075194	0.085155	0.024263	-0.148604	-0.297240	C
	2015- 11-06	145	0.6	-0.050866	-0.381870	0.036616	-0.216424	-0.402808	C