part-2

October 6, 2023

```
[]: import numpy as np
     import pandas as pd
     from matplotlib import pyplot as plt
     import cvxpy as cp
[]: raw_data = pd.read_excel('Data.xlsx', sheet_name=None, index_col=0)
[]: data = {}
     sheet_names = list(raw_data.keys())
     for s in sheet_names:
         if s == 'FamaFrenchFactors':
            new_name = s
            col_labels = raw_data[s].columns
         else:
            new_name = s[18:]
            col_labels = range(1, 11)
         data[new_name] = raw_data.pop(s)
         data[new_name].columns = col_labels
         data[new_name].index = (pd.to_datetime(data[new_name].index, format='%Y%m')
                                 .to_period('M').rename('Date'))
```

0.1 Question 3

```
class Portfolio:
    def __init__(
        self, mu: pd.Series, S: pd.DataFrame,
        long_only: bool = False,
        ret: float = None, sd: float = None, sharpe: float = None,
        weights: np.ndarray = None
) -> None:
        self.mu = mu
        self.S = S
        self.long_only = long_only
        self.ret = ret
        self.sd = sd
        self.sharpe = sharpe
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self.weights = weights
         def optimize_weights(self, target_ret: float = None) -> None:
             w = cp.Variable(self.mu.size)
             ret = self.mu.values @ w
             var = cp.quad_form(w, self.S.values)
             if target_ret is not None:
                 self.ret = target_ret
             elif self.ret is None:
                 print("Error: target return not specified")
                 return
             if self.long_only:
                 prob = cp.Problem(cp.Minimize(var),
                                   [cp.sum(w) == 1, ret == self.ret, w >= 0])
             else:
                 prob = cp.Problem(cp.Minimize(var),
                                   [cp.sum(w) == 1, ret == self.ret])
             var_opt = prob.solve()
             self.ret = float(self.ret)
             self.sd = np.sqrt(var opt)
             self.weights = np.array(w.value)
         def get_sharpe(self, risk_free_rate: float) -> float:
             self.sharpe = (self.ret - risk_free_rate) / self.sd
             return self.sharpe
[]: class P_Efficient_Frontier:
         def __init__(
             self,
             ret_range: np.ndarray,
             portfolios: list[Portfolio] = []
     ) -> None:
```

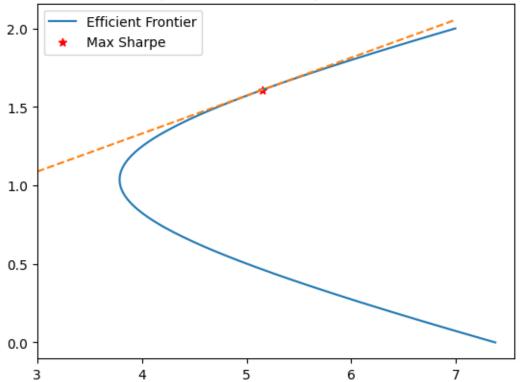
```
def __init__(
    self,
    ret_range: np.ndarray,
    portfolios: list[Portfolio] = []
) -> None:
    self.ret_range = list(ret_range)
    self.portfolios = portfolios

def generate_ef(self, mu: pd.Series, S: pd.DataFrame, long_only: bool =_u
    False):
    self.portfolios = []
    for r in self.ret_range:
        p = Portfolio(mu, S, long_only=long_only)
        p.optimize_weights(target_ret=r)
```

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self.portfolios.append(p)
        def get_tan_p(self, risk_free_rate: float) -> Portfolio:
            list_sharpe = []
             for i in range(len(self.ret_range)):
                 list_sharpe.append(self.portfolios[i].get_sharpe(risk_free_rate))
             idx_max_sharpe = np.argmax(np.array(list_sharpe))
             return self.portfolios[idx_max_sharpe]
        def get_sd_ret_arr(self):
            sd_arr = []
            ret_arr = []
             for p in self.portfolios:
                 sd_arr.append(p.sd)
                ret_arr.append(p.ret)
            return (sd_arr, ret_arr)
[]: rf mu = data['FamaFrenchFactors']['RF'].mean()
    rf_mu
[ ]: 0.36216366158113733
[]: p inv mu = data['Investment'].mean()
    p_inv_ex = data['Investment'] - p_inv_mu
    p_inv_S = p_inv_ex.cov()
[]: ef_inv = P_Efficient_Frontier(ret_range=np.linspace(0, 2, 200))
    ef_inv.generate_ef(p_inv_mu, p_inv_S)
    tanp_inv = ef_inv.get_tan_p(risk_free_rate=rf_mu)
[]: tanp_inv.ret, tanp_inv.sd, tanp_inv.sharpe
[]: (1.6080402010050252, 5.147896588073307, 0.2420166213731538)
[]: tanp_inv.weights
[]: array([0.40781556, 0.79465305, 0.52837744, -0.20908625, -0.11511586,
           -0.31282123, 0.3852204, 0.31157557, 0.86795987, -1.65857855])
[]: _, ax = plt.subplots()
    sd_ef_inv, ret_ef_inv = ef_inv.get_sd_ret_arr()
    ax.plot(sd_ef_inv, ret_ef_inv, label='Efficient Frontier')
```

```
ax.scatter(tanp_inv.sd, tanp_inv.ret, marker="*", c="r", label="Max Sharpe")
ax.set_xlim(3, )
ax.plot([3, 7],
        [3 * tanp_inv.sharpe + rf_mu, 7 * tanp_inv.sharpe + rf_mu],
        '--')
ax.set_title("10 portfolios formed by investment")
ax.legend()
plt.show()
```

10 portfolios formed by investment

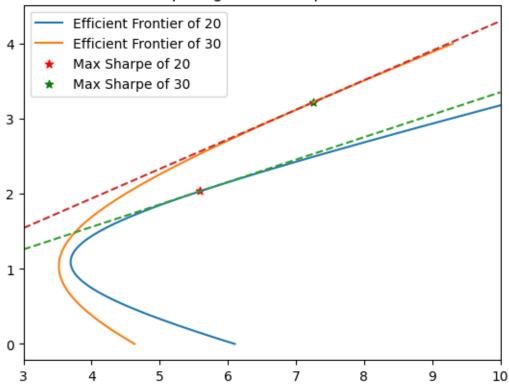


```
[]: p_20_mu = p_20.mean()
p_20_S = (p_20 - p_20_mu).cov()
```

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[]: ef_20 = P_Efficient_Frontier(ret_range=np.linspace(0, 4, 400))
    ef_20.generate_ef(p_20_mu, p_20_S)
    tanp_20 = ef_20.get_tan_p(risk_free_rate=rf_mu)
[]: tanp_20.ret, tanp_20.sd, tanp_20.sharpe
[]: (2.0350877192982457, 5.593824346035855, 0.2990662477456323)
[]: tanp_20.weights
[]: array([0.78988051, 0.99014918, 0.56167952, 0.58888532, 0.17093563,
           -0.25742778, 0.28170231, 0.31519619, 0.78938115, -0.95186396,
           -0.37081168, -0.60565736, -0.78919639, -0.54669002, 0.44796668,
           -0.75018478, -0.22358038, 0.06362291, 0.63441061, -0.13839766)
[]: p_30 = pd.concat([data['Investment'], data['Profitability'], data['Momentum']],
                     axis=1)
    p_30.columns = ([f'inv_{i}' for i in data['Investment'].columns]
                    + [f'prof_{i}' for i in data['Profitability'].columns]
                    + [f'mom {i}' for i in data['Momentum'].columns])
[]: p_30_mu = p_30.mean()
    p_30_S = (p_30 - p_30_mu).cov()
[]: ef_30 = P_Efficient_Frontier(ret_range=np.linspace(0, 4, 400))
    ef_30.generate_ef(p_30_mu, p_30_S)
    tanp_30 = ef_30.get_tan_p(risk_free_rate=rf_mu)
[]: tanp_30.ret, tanp_30.sd, tanp_30.sharpe
[]: (3.2180451127819545, 7.250488713684098, 0.3938881314042753)
[]: tanp_30.weights
[]: array([0.47510953, 0.78206432, 0.63087101, 0.36542712, -0.0805739,
           -0.36666614, -0.02724849, -0.11491686, 0.29314728, -1.49698766,
           -0.5103277, -0.55616111, -0.64796704, -0.55649414, 0.67394582,
           -0.79013437, 0.19087547, 0.39121704, 0.45853068, 0.09833374,
           -0.63605413, 0.43168549, 0.9173083, 0.78970808, -0.45781271,
           -0.04901819, -0.52625056, -0.03100014, -0.46699115, 1.81638042])
[]: _, ax = plt.subplots()
    sd_ef_20, ret_ef_20 = ef_20.get_sd_ret_arr()
    ax.plot(sd_ef_20, ret_ef_20, label='Efficient Frontier of 20')
    sd_ef_30, ret_ef_30 = ef_30.get_sd_ret_arr()
    ax.plot(sd_ef_30, ret_ef_30, label='Efficient Frontier of 30')
```

```
ax.scatter(tanp_20.sd, tanp_20.ret, marker="*", c="r", label="Max Sharpe of 20")
ax.scatter(tanp_30.sd, tanp_30.ret, marker="*", c="g", label="Max Sharpe of 30")
ax.set_xlim(3, 10)
ax.plot([3, 10],
        [3 * tanp_20.sharpe + rf_mu, 10 * tanp_20.sharpe + rf_mu],
        '--')
ax.plot([3, 10],
        [3 * tanp_30.sharpe + rf_mu, 10 * tanp_30.sharpe + rf_mu],
        '--')
ax.set_title("Comparing 20 and 30 portfolios")
ax.legend()
plt.show()
```

Comparing 20 and 30 portfolios



```
[]: df_tanp_30 = pd.DataFrame({
    'mean_return': p_30_mu,
    'sd_return': np.sqrt(np.diag(p_30_S))
```

```
})
    df_tanp_30['sharpe'] = (df_tanp_30['mean_return'] - rf_mu) /__

df_tanp_30['sd_return']

    df_tanp_30['weights'] = tanp_30.weights
    df_tanp_30 = df_tanp_30.sort_values(by='weights', ascending=False)
    df_tanp_30
[]:
             mean_return sd_return
                                        sharpe
                                                 weights
    mom_10
                 1.475284
                            6.118831
                                     0.181917
                                                1.816380
    mom_3
                 0.913426
                            5.449894
                                     0.101151 0.917308
    mom 4
                 0.942635
                            4.899604 0.118473 0.789708
    inv_2
                 1.151318
                            4.750456 0.166122 0.782064
                            4.675570 0.135971 0.673946
    prof 5
                 0.997906
    inv_3
                 1.067323
                            4.356696 0.161857 0.630871
    inv_1
                            5.358714 0.149621 0.475110
                 1.163939
                            4.547001 0.156833 0.458531
    prof_9
                 1.075284
    mom_2
                 0.774466
                            6.488377
                                     0.063545 0.431685
                                     0.147540 0.391217
                            4.526771
    prof_8
                 1.030042
    inv_4
                 0.973245
                            4.230571
                                     0.144444 0.365427
                 0.997379
                            5.392302 0.117800 0.293147
    inv_9
    prof_7
                 0.898044
                            4.600458 0.116484 0.190875
                                     0.145650 0.098334
    prof_10
                 1.038655
                            4.644635
    inv_7
                0.972663
                            4.400360
                                     0.138738 -0.027248
    mom 8
                 1.049528
                            4.456146
                                     0.154251 -0.031000
                 0.937004
    mom 6
                            4.558031 0.126116 -0.049018
    inv 5
                            4.198568 0.142959 -0.080574
                 0.962386
    inv_8
                0.979251
                            4.771906 0.129317 -0.114917
    inv 6
                            4.390712 0.134943 -0.366666
                 0.954660
    mom_5
                            4.498224 0.110748 -0.457813
                 0.860333
    mom_9
                            4.756176 0.156491 -0.466991
                 1.106463
    prof_1
                 0.748890
                            6.578407
                                     0.058787 -0.510328
    mom_7
                 0.946117
                            4.391049
                                     0.132987 -0.526251
                 0.801942
                            5.189462
                                     0.084744 -0.556161
    prof_2
    prof_4
                 0.889695
                            4.616743
                                     0.114265 -0.556494
                0.304411
                            8.392381 -0.006882 -0.636054
    mom_1
    prof_3
                 0.837323
                            4.939107
                                     0.096204 -0.647967
    prof_6
                 0.922080
                            4.554809
                                     0.122929 -0.790134
    inv_10
                0.770569
                            6.083457
                                      0.067134 -1.496988
[]: df_tanp_30.corr()
[]:
                 mean_return sd_return
                                            sharpe
                                                     weights
    mean_return
                     1.000000 -0.529972 0.928828 0.631035
    sd_return
                    -0.529972
                                1.000000 -0.756718 -0.079825
    sharpe
                     0.928828 -0.756718 1.000000 0.521095
```

