

Asset Management HW3

Ze Yang
Zhengyang Qi

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1 Problem 1

1.1 (a)

The Lagrangian of agent i 's problem is

$$\mathcal{L}_i(\mathbf{x}, \lambda) = (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0)^\top \mathbf{x} - \frac{\gamma_i}{2} \mathbf{x}^\top \boldsymbol{\Omega} \mathbf{x} - \lambda(m_i(\mathbf{P}^0)^\top \mathbf{x} - W_i) \quad (1)$$

\mathbf{x}^i, λ_i solves the KKT condition:

$$\begin{cases} \nabla_{\mathbf{x}} \mathcal{L} = (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0) - \gamma_i \boldsymbol{\Omega} \mathbf{x} - \lambda_i m_i \mathbf{P}^0 = \mathbf{0} \\ m_i(\mathbf{P}^0)^\top \mathbf{x} - W_i \leq 0 \\ \lambda_i \geq 0 \\ \lambda_i(m_i(\mathbf{P}^0)^\top \mathbf{x} - W_i) = 0 \end{cases} \quad (2)$$

The first equality yields $\mathbf{x}^i = \frac{1}{\gamma_i} \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - (1 + \lambda_i m_i) \mathbf{P}^0)$. Assume the constraint is binding, then

$$\begin{aligned} 0 &= m_i(\mathbf{P}^0)^\top \mathbf{x}^i - W_i = \frac{1}{\gamma_i} m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - (1 + \lambda_i m_i) \mathbf{P}^0) - W_i \\ \Rightarrow \frac{\lambda_i}{\gamma_i} m_i^2(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} \mathbf{P}^0 &= \frac{1}{\gamma_i} m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0) - W_i \\ \Rightarrow \lambda_i &= \frac{m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0) - \gamma_i W_i}{m_i^2(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} \mathbf{P}^0} \end{aligned} \quad (3)$$

Dual feasibility requires $\lambda_i \geq 0 \Rightarrow m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0) - \gamma_i W_i \geq 0$ (\dagger). If instead we have (\dagger) < 0 , we have $\lambda_i = 0$. Then $\mathbf{x}^i = \frac{1}{\gamma_i} \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0)$, hence $m_i(\mathbf{P}^0)^\top \mathbf{x}^i - W_i = \frac{1}{\gamma_i} m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0) - W_i < 0$ as the risk aversion coefficient $\gamma_i > 0$. The primal feasibility is satisfied. To conclude:

$$\begin{aligned} \mathbf{x}^i &= \frac{1}{\gamma_i} \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - (1 + \psi_i) \mathbf{P}^0) \\ \text{where } \psi_i &= \begin{cases} \frac{m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0) - \gamma_i W_i}{m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} \mathbf{P}^0} & \text{if } m_i(\mathbf{P}^0)^\top \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - \mathbf{P}^0) - \gamma_i W_i \geq 0 \\ 0 & \text{otherwise} \end{cases} \end{aligned} \quad (4)$$

Consequently

$$\begin{aligned} \mathbf{x}^* &= \sum_{i=1}^I \mathbf{x}_i = \sum_{i=1}^I \frac{1}{\gamma_i} \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - (1 + \psi_i) \mathbf{P}^0) \\ &= \frac{1}{\gamma} \boldsymbol{\Omega}^{-1} (\mathbb{E}[\mathbf{P}^1] - (1 + \psi) \mathbf{P}^0) \\ \text{where } \gamma &= \frac{1}{\sum_{i=1}^I 1/\gamma_i}; \quad \psi = \frac{\sum_{i=1}^I \psi_i / \gamma_i}{\sum_{i=1}^I 1/\gamma_i} \end{aligned} \quad (5)$$

1.2 (b)

Premultiply both sides of (5) by $\gamma\Omega$:

$$\begin{aligned}\gamma\Omega\mathbf{x}^* &= \mathbb{E}[\mathbf{P}^1] - (1 + \psi)\mathbf{P}^0 \\ \Rightarrow \mathbb{E}[\mathbf{P}^1] &= (1 + \psi)\mathbf{P}^0 + \gamma\Omega\mathbf{x}^* \\ \Rightarrow \frac{\mathbb{E}[P_s^1] - P_s^0}{P_s^0} &= \psi + \frac{\gamma}{P_s^0}(\Omega\mathbf{x}^*)_s\end{aligned}\tag{6}$$

Since P_s^0 is non-random, we conclude that $\mathbb{E}[r_s] = \frac{\mathbb{E}[P_s^1] - P_s^0}{P_s^0} = \psi + \frac{\gamma}{P_s^0}(\Omega\mathbf{x}^*)_s$.

1.3 (c)

By definition of \mathbf{h} , we know $\mathbf{h}^\top \mathbf{1} = 1$. Let $\mu_s := (\Omega\mathbf{x}^*)_s / P_s^0 = \sum_{j=1}^S \Omega_{sj} x_j^* / P_s^0$ for $s = 1, \dots, S$.

$$\begin{aligned}\frac{\mathbb{E}[r_M] - \psi}{\text{Var}[r_M]} &= \frac{\mathbb{E}[\mathbf{r}^\top \mathbf{h}] - \psi}{\text{Var}[\mathbf{r}^\top \mathbf{h}]} = \frac{\mathbf{h}^\top \mathbb{E}[\mathbf{r}] - \psi}{\mathbf{h}^\top \text{Cov}[\mathbf{r}] \mathbf{h}} \\ &= \frac{\mathbf{h}^\top (\psi \mathbf{1} + \gamma \boldsymbol{\mu}) - \psi}{\mathbf{h}^\top \text{Cov}[\mathbf{r}] \mathbf{h}} = \frac{\gamma \mathbf{h}^\top \boldsymbol{\mu} + \psi \overbrace{(\mathbf{h}^\top \mathbf{1} - 1)}^0}{\mathbf{h}^\top \mathbf{V} \mathbf{h}} \\ &= \frac{\gamma \sum_{i=1}^S h_i \mu_i}{\sum_{i=1}^S \sum_{j=1}^S h_i V_{ij} h_j} = \gamma \frac{\sum_{i=1}^S \left(\frac{P_i^0 x_i^*}{(\mathbf{P}^0)^\top \mathbf{x}^*} \right) \left(\sum_{j=1}^S \Omega_{ij} x_j^* \frac{1}{P_i^0} \right)}{\sum_{i=1}^S \sum_{j=1}^S \frac{P_i^0 x_i^*}{(\mathbf{P}^0)^\top \mathbf{x}^*} V_{ij} \frac{P_j^0 x_j^*}{(\mathbf{P}^0)^\top \mathbf{x}^*}} \\ &= \gamma (\mathbf{P}^0)^\top \mathbf{x}^* \frac{\sum_{i=1}^S \sum_{j=1}^S \textcolor{red}{P_i^0} \textcolor{blue}{x_i^*} \textcolor{blue}{P_i^0} V_{ij} \textcolor{blue}{P_j^0} \textcolor{blue}{x_j^*} \frac{1}{\textcolor{red}{P_i^0}}}{\sum_{i=1}^S \sum_{j=1}^S \textcolor{blue}{x_i^*} \textcolor{blue}{P_i^0} V_{ij} \textcolor{blue}{P_j^0} \textcolor{blue}{x_j^*}} \\ &= \gamma (\mathbf{P}^0)^\top \mathbf{x}^*\end{aligned}\tag{7}$$

1.4 (d)

Use the results of (c):

$$\begin{aligned}RHS &= \psi + \beta + s(\mathbb{E}[r_M] - \psi) = \psi + \text{Cov}[r_s, r_M] \frac{\mathbb{E}[r_M] - \psi}{\text{Var}[r_M]} \\ &= \psi + \text{Cov}[r_s, \sum_{j=1}^S h_j r_j] \gamma (\mathbf{P}^0)^\top \mathbf{x}^* = \psi + \gamma (\mathbf{P}^0)^\top \mathbf{x}^* \sum_{j=1}^S V_{sj} h_j \\ &= \psi + \gamma (\mathbf{P}^0)^\top \mathbf{x}^* \sum_{j=1}^S \frac{\Omega_{sj}}{P_s^0 P_j^0} \frac{P_j^0 x_j^*}{(\mathbf{P}^0)^\top \mathbf{x}^*} \\ &= \psi + \gamma \sum_{j=1}^S \frac{\Omega_{sj} x_j^*}{P_s^0} = \psi + \gamma \frac{1}{P_s^0} (\Omega\mathbf{x}^*)_s = \mathbb{E}[r_s] \quad (\text{By the result of (b)})\end{aligned}\tag{8}$$

Which completes the proof.

2 Problem 2

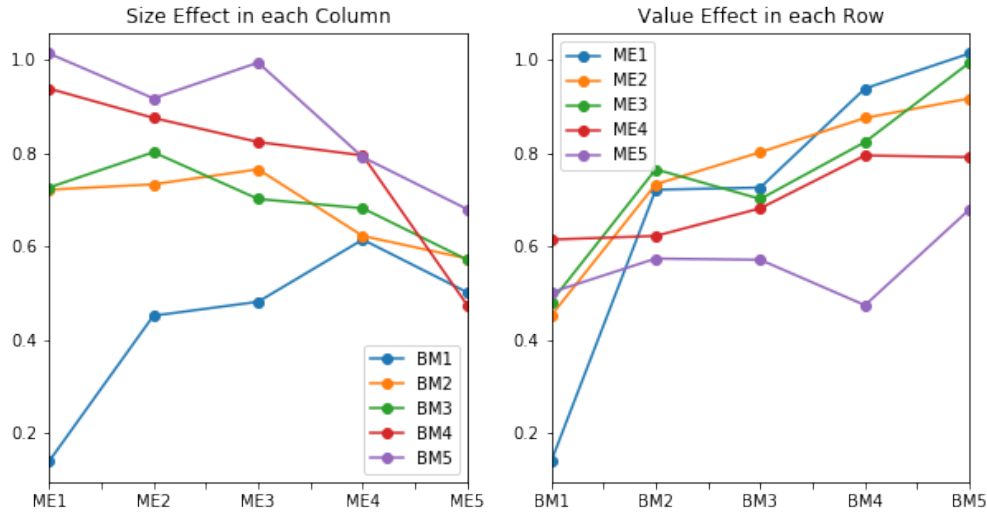
Unit for returns: percent, the same as what was used in Frenchs Data Library

2.1 (a)

Out [72]:

		BM1	BM2	BM3	BM4	BM5
ME1	mean	0.138941	0.721294	0.725931	0.938094	1.013257
	se	0.322117	0.280844	0.242438	0.229816	0.243004
	t-stat	0.431337	2.568314	2.994295	4.081942	4.169705
ME2	mean	0.451483	0.732897	0.801649	0.875069	0.916870
	se	0.293281	0.244021	0.220284	0.211684	0.244061
	t-stat	1.539423	3.003419	3.639155	4.133851	3.756730
ME3	mean	0.481265	0.765048	0.701462	0.823439	0.993619
	se	0.269862	0.221977	0.203838	0.198656	0.229668
	t-stat	1.783373	3.446511	3.441274	4.145044	4.326334
ME4	mean	0.614499	0.622239	0.681545	0.794719	0.791070
	se	0.242311	0.210575	0.202576	0.193848	0.230790
	t-stat	2.535990	2.954949	3.364389	4.099694	3.427656
ME5	mean	0.501318	0.573863	0.571117	0.473859	0.679433
	se	0.189887	0.181751	0.175723	0.189212	0.219410
	t-stat	2.640088	3.157413	3.250092	2.504379	3.096636

2.2 (b)



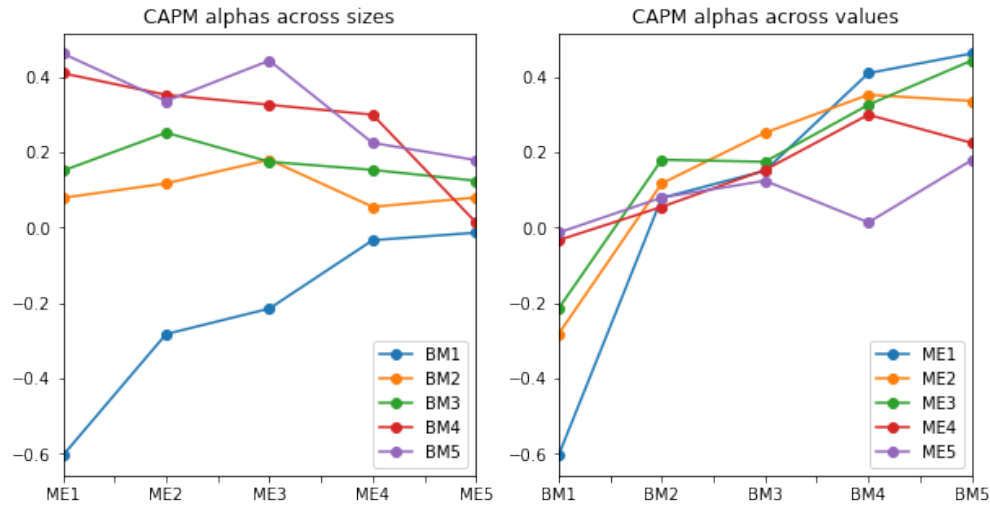
- For the same BM quantile group (above BM2), we can observe a size effect across different size groups, in the sense that the excess return increases as size group goes from BIG to SMALL. (Figure 1)
- The size effect described above is not evident for LoBM(BM1) group: the excess returns do not increase monotonically.
- For the same size quantile group, we can observe a value effect across different BM groups, in the sense that the excess return increases as BM group goes from LowBM to HighBM. (Figure 2)
- The value effect is stronger in the SMALL(ME1) size group, as the blue line in figure 2 has the steepest slope.

2.3 (c)

Out [69]:

		BM1	BM2	BM3	BM4	BM5
ME1	alpha	-0.604325	0.0792367	0.151235	0.410014	0.462256
	t-stat	-3.08398	0.456519	1.079	2.91167	3.03706
	p<0.05	(*)			(*)	(*)
ME2	alpha	-0.28214	0.117712	0.252293	0.352649	0.336371
	t-stat					

t-stat	-1.8987	0.975059	2.24046	3.1656	2.40059
p<0.05			(*)	(*)	(*)
ME3 alpha	-0.214433	0.180408	0.17511	0.326279	0.443554
t-stat	-1.73615	1.94766	1.88901	3.24615	3.41398
p<0.05				(*)	(*)
ME4 alpha	-0.033038	0.0555156	0.153521	0.299895	0.225168
t-stat	-0.352906	0.71254	1.73193	3.25954	1.82287
p<0.05				(*)	
ME5 alpha	-0.0130529	0.0802354	0.124674	0.0141237	0.179359
t-stat	-0.194115	1.27065	1.47073	0.136355	1.31604
p<0.05					



2.4 (d)

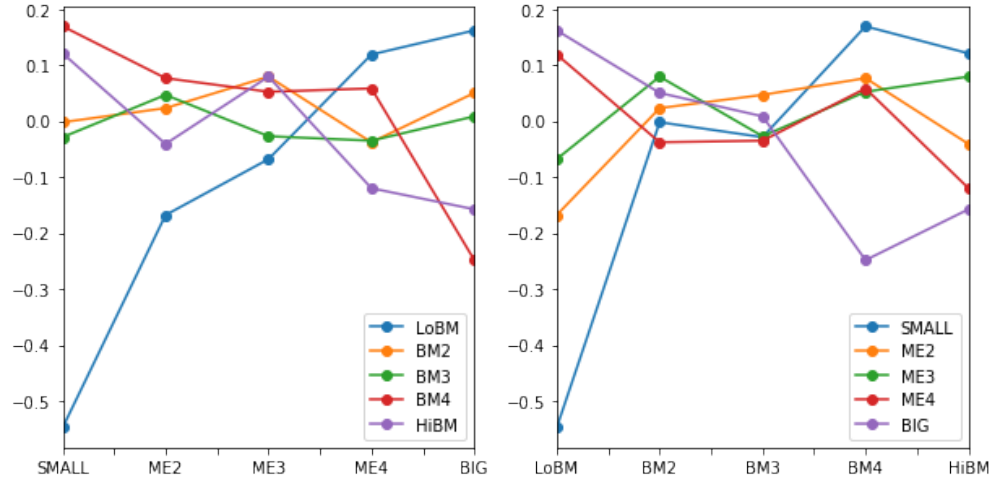
Out [68]:

		BM1	BM2	BM3	BM4	BM5
ME1	alpha	-0.547419	-0.00188453	-0.0285962	0.169426	0.121075
	t-stat	-5.62311	-0.0260974	-0.523543	3.10497	2.11213
	p<0.05				(*)	(*)
ME2	alpha	-0.167721	0.0232988	0.0469446	0.0768339	-0.0407018
	t-stat	-2.46079	0.395727	0.793127	1.48421	-0.733885
	p<0.05		(*)			
ME3	alpha	-0.0678117	0.0795249	-0.0269168	0.0524966	0.079584
	t-stat	-1.09019	1.19467	-0.405653	0.817183	1.015
	p<0.05					
ME4	alpha	0.11902	-0.0376912	-0.0348759	0.0584575	-0.119813
	t-stat	1.91318	-0.521096	-0.468031	0.859294	-1.39163
	p<0.05					
ME5	alpha	0.162147	0.0504822	0.00859773	-0.248215	-0.157442
	t-stat	3.50773	0.881661	0.123204	-3.80089	-1.57558
	p<0.05				(*)	

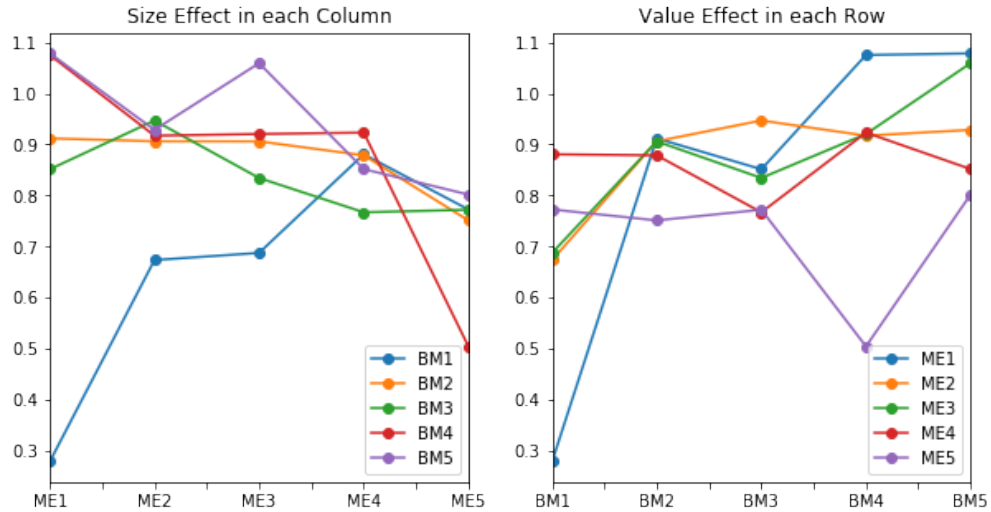
Comments:

- The alphas that are statistically significant in part (c), i.e. the CAPM single factor model becomes either
 - Much smaller in scale, in terms of the absolute values. When we inspect the plots of alphas across different groups, we can see a much messy pattern as opposed to the CAPM alphas. OR

- Becomes not statistically significant. The (*) in table marks those alphas with p-value < 0.05 . We can observe that many alphas in the top-right corner (small size, high B/M) that are previously significant now becomes not significant to the confidence level of 0.95.
- These observations bespeaks the fact that a considerable amount of the CAPM alphas on the size \times value quantile portfolios can be explained by the SMB and HML risk factors.



2.5 (e)



Repeat Part (a)

	BM1	BM2	BM3	BM4	BM5
ME1 mean	0.279520	0.911910	0.851436	1.074964	1.078133
se	0.397315	0.346695	0.281501	0.270598	0.282844
t-stat	0.703521	2.630298	3.024629	3.972553	3.811763
ME2 mean	0.673830	0.906238	0.946844	0.917159	0.928510
se	0.350828	0.284240	0.255377	0.254190	0.302355
t-stat	1.920685	3.188284	3.707637	3.608169	3.070930
ME3 mean	0.687848	0.906062	0.834394	0.920435	1.059032
se	0.323769	0.259975	0.240980	0.243974	0.278055
t-stat	2.124506	3.485189	3.462497	3.772677	3.808714

ME4 mean	0.880749	0.878720	0.767234	0.923185	0.851772
se	0.291348	0.238622	0.247059	0.232234	0.283599
t-stat	3.023010	3.682481	3.105468	3.975237	3.003433
ME5 mean	0.772807	0.751408	0.772299	0.505007	0.801949
se	0.221474	0.210062	0.208792	0.244736	0.295633
t-stat	3.489385	3.577085	3.698892	2.063480	2.712648

Repeat Part (b): see the plots above.

Repeat Part (c)

		BM1	BM2	BM3	BM4	BM5
ME1 alpha	-0.693204	0.0743478	0.127975	0.41341	0.376434	
t-stat	-2.5771	0.311648	0.719195	2.25265	1.99988	
p<0.05	(*)			(*)	(*)	
ME2 alpha	-0.27519	0.12434	0.254925	0.235765	0.140363	
t-stat	-1.36331	0.789451	1.74087	1.58208	0.751792	
p<0.05						
ME3 alpha	-0.214548	0.147307	0.146577	0.256493	0.327003	
t-stat	-1.23544	1.22032	1.21334	1.8511	1.93806	
p<0.05						
ME4 alpha	0.0251955	0.170176	0.0642887	0.267753	0.0885353	
t-stat	0.190777	1.64981	0.514063	2.22401	0.53679	
p<0.05				(*)		
ME5 alpha	0.0923624	0.11796	0.190769	-0.144086	0.0422177	
t-stat	1.17026	1.40275	1.70035	-0.98358	0.225891	
p<0.05						

Repeat Part (d)

		BM1	BM2	BM3	BM4	BM5
ME1 alpha	-0.634844	0.047471	0.0315884	0.270725	0.175225	
t-stat	-4.60346	0.448571	0.424996	3.60445	2.33627	
p<0.05	(*)			(*)	(*)	
ME2 alpha	-0.203053	0.0659427	0.119476	0.0594593	-0.0984468	
t-stat	-2.24812	0.837165	1.53178	0.879905	-1.3781	
p<0.05	(*)					
ME3 alpha	-0.108831	0.0794007	0.0143403	0.078121	0.101301	
t-stat	-1.3034	0.873981	0.162489	0.856674	0.945184	
p<0.05						
ME4 alpha	0.11877	0.0891055	-0.0694418	0.124072	-0.125605	
t-stat	1.43847	0.965733	-0.680546	1.353	-1.07457	
p<0.05						
ME5 alpha	0.189886	0.079747	0.0923893	-0.3241	-0.172125	
t-stat	3.71963	1.11945	1.1159	-3.70759	-1.28333	
p<0.05	(*)			(*)		

We used 1988 - 2018, i.e. recent 30 years for this section. **Differences:**

- In the recent period, the excess return profile across size× value double sorts becomes much more messy. As in the plots we can't find the downward/upward sloping average returns.
- Less portfolios from the double sorts have statistically significant CAPM alphas and FF3 alphas: the market becomes more efficient.

Similarities:

- The (Small size \times high BM) portfolios in the top-right corner still has significant CAPM alphas, and that alpha reduces in a similar fashion as we account for FF3 factors.

3 Problem 3

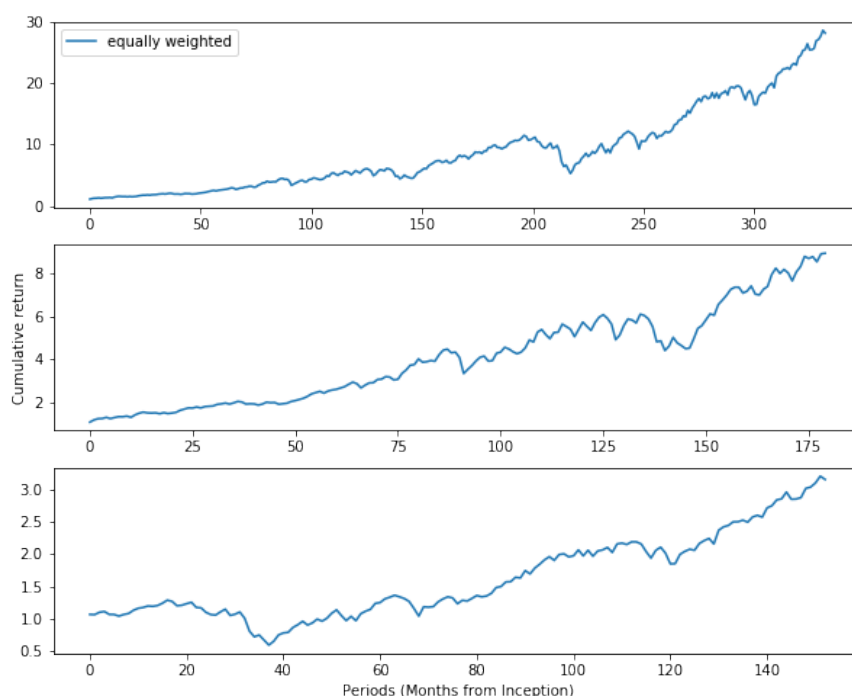
Note: all the sharpe ratios are annualized, i.e. scaled by multiplying $\sqrt{12}$

3.1 (a)

Sharpe ratio (1991/01 - 2018/10): 0.6705

Sharpe ratio (1991/01 - 2005/12): 0.7862

Sharpe ratio (2006/01 - 2018/10): 0.5523

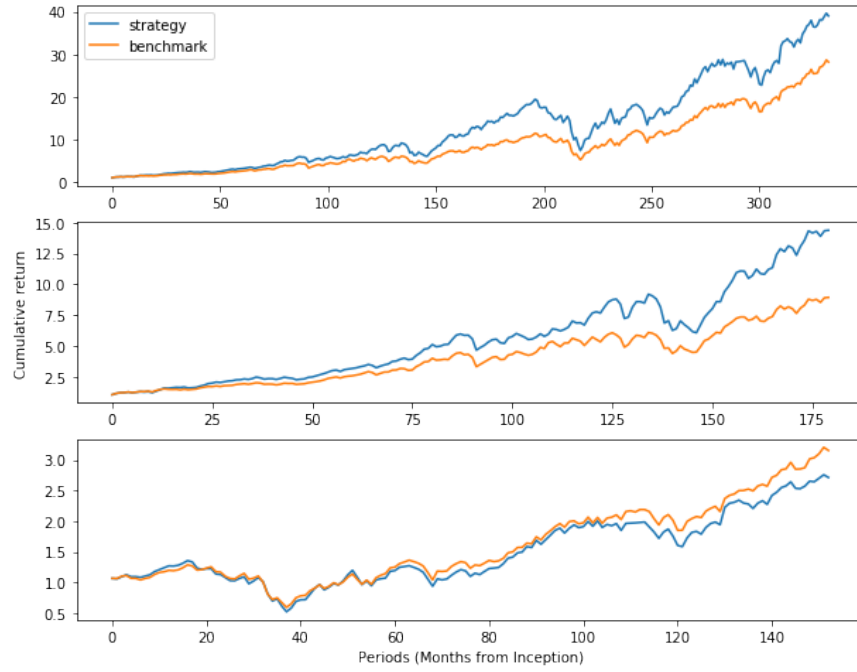


3.2 (b)

Sharpe ratio (1991/01 - 2018/10): 0.6916

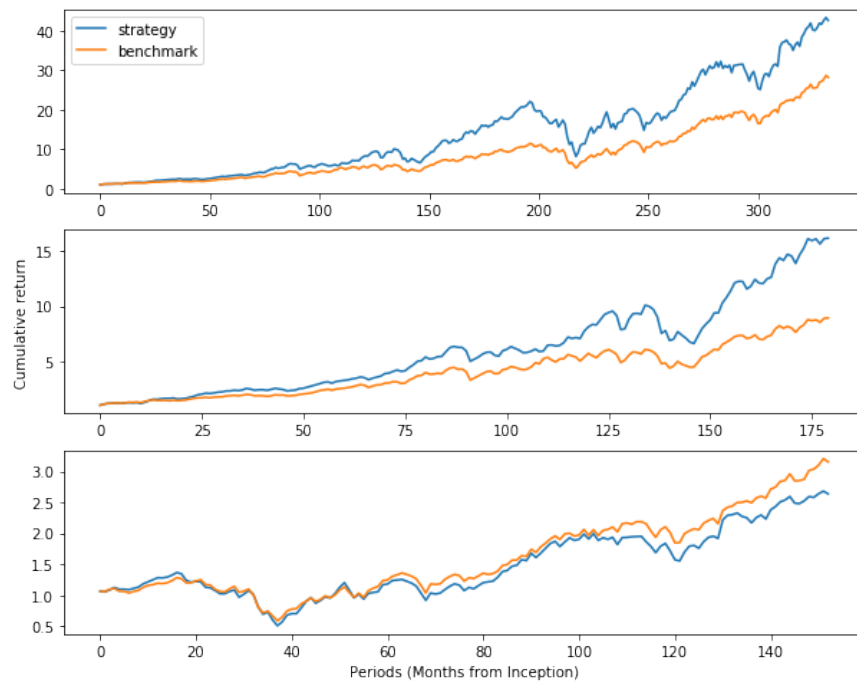
Sharpe ratio (1991/01 - 2005/12): 0.9773

Sharpe ratio (2006/01 - 2018/10): 0.4414



3.3 (c)

Sharpe ratio (1991/01 - 2018/10): 0.7013
 Sharpe ratio (1991/01 - 2005/12): 1.0266
 Sharpe ratio (2006/01 - 2018/10): 0.4248

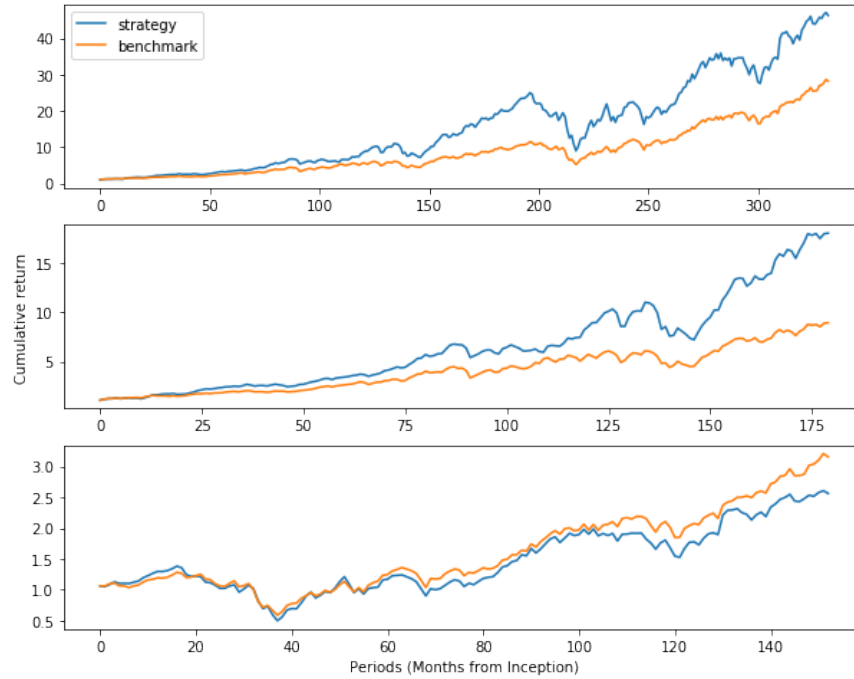


3.4 (d)

Sharpe ratio (1991/01 - 2018/10): 0.7074

Sharpe ratio (1991/01 - 2005/12): 1.0659

Sharpe ratio (2006/01 - 2018/10): 0.4083



3.5 (e)

Comments:

- In terms of aggressiveness, we have $(d) > (c) > (b) > (a)$.
- More aggressive strategy has higher return and sharpe ratio in 1991-2018 (“overall”), and 1991-2005 (“good periods”); and outperforms the benchmark in those periods.
- More aggressive strategy has lower return and sharpe ratio in 2006-2018 (“bad/high volatility periods”); and underperforms the benchmark in those periods.

Limitations:

- The margin requirements ask for large amount of capital to implement more aggressive strategies.
- Short selling may not be easy for some countries/exchanges.
- The aggressive strategies have higher drawdowns in the “bad” periods, which may cause the fund to blow up.