## THEORY OF FINANCE

#### Solution Sheet on Problem Set 3

## **Asset Pricing Models & Portfolio Choice**

Deadline: 30.11.2021

## Solved by: Jonas Husmann, Niklas Kampe, Cyril Janak

Task		Points Earne d
1. Analyzing Beta Sorted Portfolios a) Annualized portfolio statistics (4 points)	Index	
b) CAPM regression & report of statistics (8 points)	Index	
c) Plot and interpret returns and betas (8 points)	Problem 1 - Part 1 - Task c)  O.15  Security Market Line Market Portfolio  O.14  O.11  O.10  O.12  O.11  O.10  Estimated Betas  Intercept (Risk Free Rate): 0.02295  Slope: 0.09839	

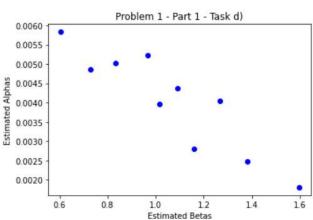
The security market line distinguishes between Portfolios above and under the line. Portfolios above the line are outperforming the market, whereas Portfolios under the line are underperforming the market. The market portfolio is equal to  $\beta = 1$ . For our dataset, we observe that the assets with a  $\beta > 1$  are underperforming, whereas the  $\beta < 1$  are outperforming the market. Recalling that a  $\beta > 1$  means that the stock's price swings more wildly (i.e. is more volatile) than the overall market, we can conclude, that the higher  $\beta$  do not perform strong enough, compared to the market (i.e. the returns are not high enough, given the higher volatility).

This is in line with our results in part a). Recall the formula for the Sharpe Ratio:

Sharpe Ratio 
$$SR = \frac{R_p - R_f}{\sigma_p}$$

The analysis in part a) gave us a decreasing Sharpe Ratio from  $\beta 1$  to  $\beta 10$ . When we look into the parameters, which are defining the Sharpe Ratio, we see that the Returns are – more or less – increasing from  $\beta 1$  to  $\beta 10$ . However, the volatility (i.e. the Standard Deviation) is increasing much more, in relation to the increasing return. This leads to the decrease in the Sharpe Ratio, which is, eventually, why they are underperforming the market.

d)
Plot and
interpret
alphas and
betas
(8 points)



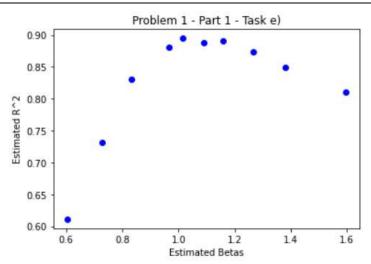
We compare how volatile a stock's price is in comparison to the overall stock market ( $\beta$ , x-axis) versus the investment strategy's ability to beat the market ( $\alpha$ , y-axis). Also called "excess return" or "abnormal rate of return".

 $\alpha$  is created by active investing, where as  $\beta$  can be earned through passive investing.

Our plot shows a decreasing relationship for  $\alpha$  and  $\beta$ . Meaning,  $\beta$ 1 has a – relatively – high abnormal return, while having a low volatility (in comparison with the market).

Based on our dataset, the lower CAPM- $\beta$  have earned higher abnormal returns, while having lower volatility than the market. On the other side, the higher CAPM- $\beta$  have earned clearly lower abnormal returns, while having much higher volatility (again, in comparison with the market). Recall that in CAPM with  $\beta$  close to zero, the return should also be close to zero. When  $\beta$  are low, the return can be explained by active investing (i.e. high  $\alpha$ ). On the other side, when  $\beta$  are high, returns can be explained by the higher volatility (i.e. higher  $\beta$ ), which is a risk premium for pro cyclical assets.

e) Plot and interpret R squared and betas (8 points)

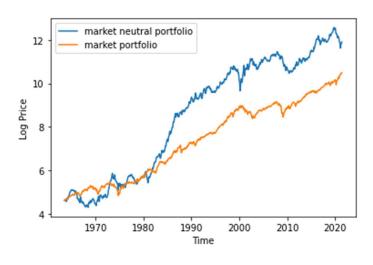


Lastly, we compare again our  $\beta$  on the x-axis with the Estimated  $R^2$ , the measure of how well observed outcomes are replicated by the model, on the y-axis.

Obviously, the higher our  $R^2$ , the more accurate the linear relation in our CAPM-Regression. We observe an increase in  $R^2$  up to  $\beta=1$  and then a (weaker compared to the increase) decreases in higher  $\beta$ . Our dataset basically shows that for high/low Beta, the linear relation between returns and market returns is weaker, compared to medium- $\beta$ . Which means, that our model is less accurate, when we analyze high/low  $\beta$  Portfolios.

f)
Build betaneutral
portfolio and
plot results
vs. market
(6 points)

The required weight for the long position in beta 1 is 2.646.



g)
Performanc
e
comparison
of betaneutral
portfolio to
market
(6 points)

Monthly results:

market neutral portfolio mean return: 0.0136 market portfolio mean return: 0.0095

market neutral portfolio sharpe ratio: 0.1251 market portfolio sharpe ratio: 0.1309

Correlation Matrix	returns_mkt_neutral	market_return
returns_mkt_neutral	1.000000	0.006087
market_return	0.006087	1.000000

Per the above correlation matrix the market neutral portfolio is indeed still correlated to the market returns, however only very slightly. Moreover, given we only rebalance the portfolio once and not e.g. every month the low correlation in this case can be considered to be market neutral.

regressions on beta and Fama-French models (14 points)

# For the following regressions: x1 = Mkt\_RF

x2 = SMB

x3 = HML

x4 = RMW

x5 = CMA

## **CAPM** regression:

Dep. Variable:					У	R-sa	uared:			0.00
Model:					OLS		R-squar	ed:		-0.00
Method:		Le	east	Squ		-	atistic:			0.00720
Date:							(F-stat	istic):		0.93
Time:		•		22:5	2:57	Log-	Likeliho	od:		776.2
No. Observation	ns:				696	AIC:				-1549
Df Residuals:					694	BIC:				-1539
Df Model:					1					
Covariance Type	e:		r	nonro	bust					
					====					
										0.975
const										0.01
x1	0.0057	7	0.	068		0.085	0.9	32	-0.127	0.13
======== Omnibus:				==== 35	.042	Durb	====== in-Watso	====== n:		1.91
Prob(Omnibus):					.000		ue-Bera			101.56
Skew:						Prob		(/-		8.80e-2

## Fama-French 3-factor regression:

OLS	Regression	Results
-----	------------	---------

Dep. Variabl Model: Method: Date: Time: No. Observat Df Residuals Df Model:	Mor	y OLS Least Squares a, 29 Nov 2021 22:52:57 696 692	Adj. R-sq F-statist Prob (F-s Log-Likel AIC:	uared: ic: tatistic)	:	0.332 0.329 114.5 3.15e-60 916.58 -1825. -1807.
	ype:	nonrobust				
	coef	std err	t	P> t	[0.025	0.975]
x1	0.3886 -1.4095	0.003 0.059 0.085 -:	6.566 16.581	0.000	0.272 -1.576	0.505 -1.243
Omnibus: Prob(Omnibus Skew: Kurtosis:	):	24.117 0.000 0.047 4.419	Jarque-Be Prob(JB):	ra (JB):		2.014 58.622 1.86e-13 37.1

#### Fama-French 5-factor regression

Dep. Variable	e:		y R	-sau	ared:		0.384
Model:					R-squared:		0.380
Method:		Least Squa					86.1
Date:					(F-statistic):		2.42e-70
Time:					ikelihood:		945.08
No. Observat:	ions:		696 A				-1878
Df Residuals	:		690 B	IC:			-1851
Df Model:			5				
Covariance Ty	ype:	nonrok	oust				
					P> t		
					0.070		
x1					0.000		
x2	-1.2419	0.086	-14.4	19	0.000	-1.411	-1.073
x3	0.3060	0.112	2.7	24	0.007	0.085	0.52
x4	0.7058	0.119	5.9	23	0.000	0.472	0.940
x5	1.0264	0.173	5.9	46	0.000		
======== Omnibus:		6.	622 D	urbi	n-Watson:	======	2.008
Prob(Omnibus	):				e-Bera (JB):		8.870
Skew:	, -		057 P				0.0119
Kurtosis:					No.		82.4

In the CAPM regression we get a very high p value for the beta coefficient (and also a non-existent  $R^2$ ). Keeping in mind that the portfolio that was used for this regression was specifically created to be market neutral (i.e. beta = 0) this does not surprise and is reasonable.

With the Fama-French 3-factor and 5-factor regression models the excess return of the market portfolio can be better described (adj  $R^2$  of 0.329 and 0.380, respectively). It can also be seen that the goodness of fit of the model increases by adding the two additional independent variables of RMW and CMA.

Moreover, irrespective whether the 3-factor or 5-factor model was used, the p values for all independent variables are extremely low, indicating that they are all significant and have an influence on the excess return of the market neutral portfolio.

For both the 3-factor and 5-factor model we get very low constants (alpha) implying that almost all the return can be explained by the 3-factors or 5-factors respectively.

We tested for multicollinearity using the VIF. Given that the VIF is below for 2.4 for all independent variables multicollinearity is not a problem in this case. Considering how the factors are chosen and created this is sensible.

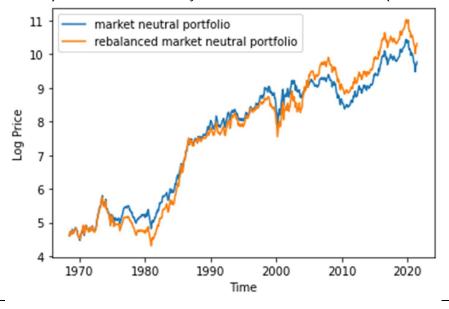
Performanc e analysis of rebalanced portfolio (10 points)

## French-Fama 5-factor regression on the monthly rebalanced portfolio:

		OLS Reg	ression F	esults		
Dep. Variable:			y R-so	uared:		0.341
Model:		0	LS Adj.	R-squared:		0.336
Method:		Least Squar	es F-st	atistic:		65.22
Date:		, 29 Nov 20		(F-statistic)	:	7.29e-55
Time:		•		Likelihood:		798.12
No. Observation	ns:		36 AIC:			-1584.
Df Residuals:		-	30 BIC:			-1558.
Df Model:			5			10001
Covariance Typ		nonrobu	7			
	.======		=======			
	coef	std err	t	P> t	10 025	0.975]
	COEI	scu err			[0.023	0.975]
const	0.0061	0.003	2.106	0.036	0.000	0.012
x1	0.4761	0.068	6.952	0.000	0.342	0.611
x2	-1.2620	0.100	-12.588	0.000	-1.459	-1.065
x3	0.2630	0.128	2.061	0.040	0.012	0.513
×4	0.6319		4.661		0.366	
x5	1.1176	0.201	5.550	0.000	0.722	1.513
Omnibus:		8.4	12 Durk	in-Watson:		1.935
Prob(Omnibus):		0.0	15 Jaro	ue-Bera (JB):		11.159
Skew:		0.1	25 Prob	(JB):		0.00377
Kurtosis:		3.5		. No.		82.0

With monthly rebalancing we get a fairly similar outcome vs if we had just done an initial weighting (cf. question 1 h). Some slight shifts in the coefficients of the independent variables can be observed, and the adjusted  $R^2$  is slightly lower here.

For further comparison we also plotted the log price development of market neutral portfolio (from 1 f) versus the log price development of the monthly rebalanced market neutral portfolio.



#### 2. Factor Rotation a) Formulate and solve the optimization function (8 points)

$$max E[r_p] + (1 - \gamma) * \frac{1}{2} * Var(r_p)$$
 s.t.  $1' * v + v_{rf} = 1$  (i)

 $\rightarrow v$  = 4x1 weight vector on risky factor portfolios,  $v_{rf}$  = weight on risk-free asset,  $r_p$  = log-return of portfolio of risky factor portfolios and risk-free asset

$$E[r_p] = v * \mu_{risky} + v_{rf} * \mu_{rf}$$
 (ii)

$$Var(r_p) = v' * \Sigma * v$$
 (iii)

 $\Rightarrow$   $\mu_{risky}$ ,  $\mu_{rf}$  = mean log returns over 2-year time horizon,  $\Sigma$  = variance-covariance matrix

(ii) and (iii) into (i):

$$\max v * \mu_{risky} + v_{rf} * \mu_{rf} + (1 - \gamma) * \frac{1}{2} * v' * \Sigma * v$$
 s.t.  $1' * v + v_{rf} = 1$ 

Lagrange formulization:

$$L = v * \mu_{risky} + v_{rf} * \mu_{rf} + (1 - \gamma) * \frac{1}{2} * v' * \Sigma * v + \lambda * [1' * v + v_{rf} - 1]$$

FOCs:

$$\frac{\delta L}{\delta v}: \ \mu^e + \frac{1}{2}*diag(\Sigma) - \gamma*\Sigma*v + \lambda = 0_{\rm nx1} \quad \text{(from 14.9)} \ \ \text{(iv)}$$

$$\frac{\delta L}{\delta v_{rf}}$$
:  $\mu_{rf} + \lambda = 0$  (v)

$$\frac{\delta L}{\delta \lambda}$$
:  $1' * v + v_{rf} - 1 = 0$  (vi)

Solving for Lambda:

$$\lambda = -\mu_{rf}$$
 (vii)

(vii) into (iv):

$$\mu^{e} + \frac{1}{2} * diag(\Sigma) - \gamma * \Sigma * v - \mu_{rf} = 0_{nx1}$$

$$\gamma * \Sigma * v = \mu^e + \frac{1}{2} * diag(\Sigma) - \mu_{rf}$$

$$\mathbf{v} = \frac{1}{\gamma} * \Sigma^{-1} * \left( \mu^e + \frac{1}{2} * diag(\Sigma) - \mu_{rf} \right)$$
 (viii)  $\rightarrow$  Optimal risky weights

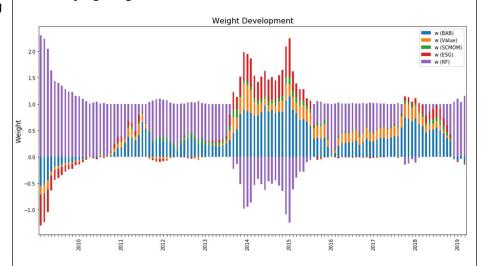
(viii) into (vi):

$$1' * \frac{1}{\gamma} * \Sigma^{-1} * \left(\mu^e + \frac{1}{2} * diag(\Sigma) - \mu_{rf}\right) + v_{rf} - 1 = 0$$

$$v_{rf} = 1 - 1' * \frac{1}{\gamma} * \Sigma^{-1} * \left(\mu^e + \frac{1}{2} * diag(\Sigma) - \mu_{rf}\right) \quad \text{(iv)} \quad \textbf{$\Rightarrow$} \quad \text{Optimal risk-free weight}$$

b) Calculate time-varying weights of optimal portfolio (10 points)

#### Time-varying weights of factor model:



based on the optimization solution:

risky weight vector: 
$$\mathbf{v} = \frac{1}{\gamma} * \Sigma^{-1} * \left( \mu^e + \frac{1}{2} * diag(\Sigma) - \mu_{rf} \right)$$
 risk-free weight:  $v_{rf} = 1 - 1' * \frac{1}{\gamma} * \Sigma^{-1} * \left( \mu^e + \frac{1}{2} * diag(\Sigma) - \mu_{rf} \right)$ 

c)
Report
performance
measures
and run
FamaFrench 5factor
regression
(10 points)

#### Performance measures:

Mean Return: 0.45% Volatility: 0.86% Sharpe Ratio: 46.3%

#### Fama-French 5-factor regression:

0LS	Reg	ression	Resu	lts

Dep. Variab	le:		y R-squa	red:		0.068	
Model:		-	OLS Adj. F	R-squared:		0.028	
Method: Date: Time: No. Observations:		Least Squa	res F-stat				
		Tue, 30 Nov 2	021 Prob (	F-statistic	):	0.142	
		20:32	:20 Log-Li	Log-Likelihood:			
			122 AIC:			-810.2	
Df Residuals:			116 BIC:			-793.3	
Df Model:			5				
Covariance	Type:	nonrob	ust				
	coef	std err	t	P> t	[0.025	0.975]	
const	0.0043	0.001	5.168	0.000	0.003	0.006	
x1	-0.0340	0.022	-1.580	0.117	-0.077	0.009	
x2	0.0596	0.039	1.543	0.125	-0.017	0.136	
x3	0.0111	0.034	0.321	0.748	-0.057	0.079	
x4	0.0388	0.052	0.749	0.455	-0.064	0.141	
x5	-0.1380	0.060	-2.307	0.023	-0.256	-0.020	
Omnibus:	======	41.	======== 547 Durbir	======== n-Watson:	=======	1.533	
Prob(Omnibu	s):	0.	000 Jarque	-Bera (JB):		90.741	
Skew:		1.	394 Prob(J	IB):		1.98e-20	
Kurtosis:		6.	174 Cond.	No.		82.7	

Based on the regression results, we can conclude that the degree of explanation of the portfolio returns, based on the Fama-French 5-factor model, is very low which can be seen at the R-squared of 6.8% and the p-value of 0.142, while the latter is normally significant when assuming confidence levels between 1-10%. Thus, the model can't explain most of the variance in the portfolio returns and leads to insignificant and unreliable results.