

THEORY OF FINANCE

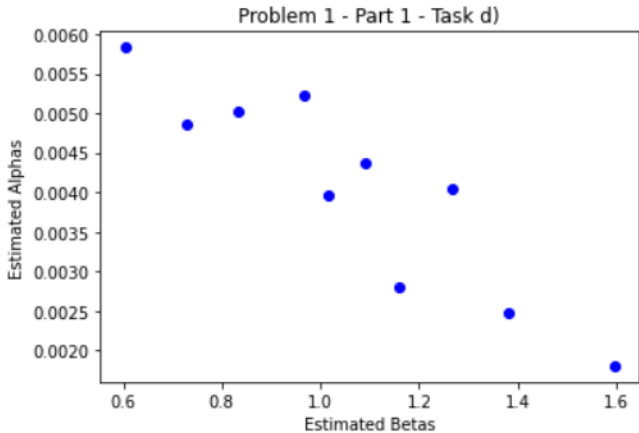
Solution Sheet on Problem Set 3

Asset Pricing Models & Portfolio Choice

Deadline: 30.11.2021

Solved by: Jonas Husmann, Niklas Kampe, Cyril Janak

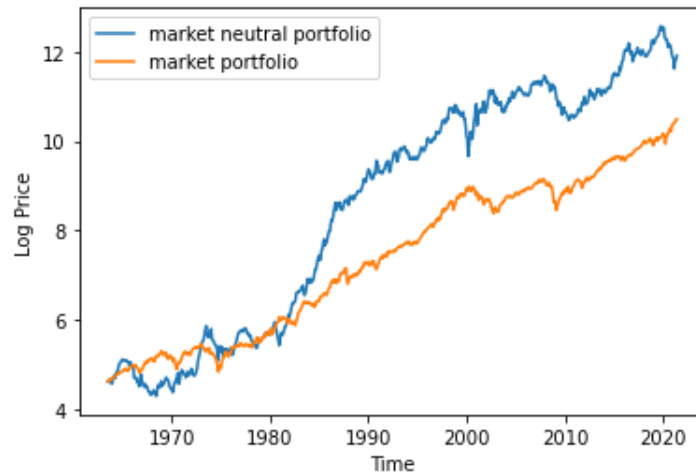
Task		Points Earned																																																																																																																									
1. Analyzing Beta Sorted Portfolios																																																																																																																											
a) Annualized portfolio statistics (4 points)	<table><tr><th>Index</th><th>beta1</th><th>beta2</th><th>beta3</th><th>beta4</th><th>beta5</th><th>beta6</th><th>beta7</th><th>beta8</th><th>beta9</th><th>beta10</th></tr><tr><td>Annual_Mean_Return</td><td>0.112186</td><td>0.109293</td><td>0.118481</td><td>0.130431</td><td>0.118619</td><td>0.128726</td><td>0.114752</td><td>0.137202</td><td>0.126393</td><td>0.133314</td></tr><tr><td>Annual_Std</td><td>0.119012</td><td>0.131352</td><td>0.140925</td><td>0.158933</td><td>0.165454</td><td>0.178226</td><td>0.189436</td><td>0.208976</td><td>0.23108</td><td>0.273389</td></tr><tr><td>Annual_Excess_Return</td><td>0.0702929</td><td>0.0672964</td><td>0.0768125</td><td>0.0891893</td><td>0.0769554</td><td>0.0874232</td><td>0.07295</td><td>0.0962018</td><td>0.0850071</td><td>0.092175</td></tr><tr><td>Annual_Excess_Std</td><td>0.119012</td><td>0.131352</td><td>0.140925</td><td>0.158933</td><td>0.165454</td><td>0.178226</td><td>0.189436</td><td>0.208976</td><td>0.23108</td><td>0.273389</td></tr><tr><td>Annual_Sharpe_Ratio</td><td>0.590637</td><td>0.512338</td><td>0.545061</td><td>0.561174</td><td>0.465116</td><td>0.490519</td><td>0.385091</td><td>0.460349</td><td>0.367868</td><td>0.337157</td></tr></table>	Index	beta1	beta2	beta3	beta4	beta5	beta6	beta7	beta8	beta9	beta10	Annual_Mean_Return	0.112186	0.109293	0.118481	0.130431	0.118619	0.128726	0.114752	0.137202	0.126393	0.133314	Annual_Std	0.119012	0.131352	0.140925	0.158933	0.165454	0.178226	0.189436	0.208976	0.23108	0.273389	Annual_Excess_Return	0.0702929	0.0672964	0.0768125	0.0891893	0.0769554	0.0874232	0.07295	0.0962018	0.0850071	0.092175	Annual_Excess_Std	0.119012	0.131352	0.140925	0.158933	0.165454	0.178226	0.189436	0.208976	0.23108	0.273389	Annual_Sharpe_Ratio	0.590637	0.512338	0.545061	0.561174	0.465116	0.490519	0.385091	0.460349	0.367868	0.337157																																																								
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c) Plot and interpret returns and betas (8 points)	<p>Problem 1 - Part 1 - Task c)</p> <p>Intercept (Risk Free Rate): 0.02295 Slope: 0.09839</p>																																																																																																																										

	<p>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 β do not perform strong enough, compared to the market (i.e. the returns are not high enough, given the higher volatility).</p> <p>This is in line with our results in part a). Recall the formula for the Sharpe Ratio:</p> $Sharpe\ Ratio\ SR = \frac{R_p - R_f}{\sigma_p}$ <p>The analysis in part a) gave us a decreasing Sharpe Ratio from β_1 to β_{10}. When we look into the parameters, which are defining the Sharpe Ratio, we see that the Returns are – more or less – increasing from β_1 to β_{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.</p>																							
d) Plot and interpret alphas and betas (8 points)	 <table><caption>Data points from the scatter plot</caption><tr><th>Estimated Betas</th><th>Estimated Alphas</th></tr><tr><td>0.6</td><td>0.0058</td></tr><tr><td>0.7</td><td>0.0048</td></tr><tr><td>0.8</td><td>0.0050</td></tr><tr><td>0.9</td><td>0.0052</td></tr><tr><td>1.0</td><td>0.0040</td></tr><tr><td>1.1</td><td>0.0044</td></tr><tr><td>1.2</td><td>0.0028</td></tr><tr><td>1.3</td><td>0.0041</td></tr><tr><td>1.4</td><td>0.0025</td></tr><tr><td>1.6</td><td>0.0018</td></tr></table> <p>We compare how volatile a stock's price is in comparison to the overall stock market (β, x-axis) versus the investment strategy's ability to beat the market (α, y-axis). Also called "excess return" or "abnormal rate of return".</p> <p>α is created by active investing, where as β can be earned through passive investing.</p>	Estimated Betas	Estimated Alphas	0.6	0.0058	0.7	0.0048	0.8	0.0050	0.9	0.0052	1.0	0.0040	1.1	0.0044	1.2	0.0028	1.3	0.0041	1.4	0.0025	1.6	0.0018	
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	<p>Our plot shows a decreasing relationship for α and β. Meaning, β_1 has a – relatively – high abnormal return, while having a low volatility (in comparison with the market).</p> <p>Based on our dataset, the lower CAPM-β have earned higher abnormal returns, while having lower volatility than the market. On the other side, the higher CAPM- β have earned clearly lower abnormal returns, while having much higher volatility (again, in comparison with the market). Recall that in CAPM with β close to zero, the return should also be close to zero. When β are low, the return can be explained by active investing (i.e. high α). On the other side, when β are high, returns can be explained by the higher volatility (i.e. higher β), which is a risk premium for pro cyclical assets.</p>																							
e) Plot and interpret R squared and betas (8 points)	<div><p>Problem 1 - Part 1 - Task e)</p><table><thead><tr><th>Estimated Betas</th><th>Estimated R^2</th></tr></thead><tbody><tr><td>0.6</td><td>0.61</td></tr><tr><td>0.7</td><td>0.73</td></tr><tr><td>0.8</td><td>0.83</td></tr><tr><td>0.9</td><td>0.88</td></tr><tr><td>1.0</td><td>0.89</td></tr><tr><td>1.1</td><td>0.88</td></tr><tr><td>1.2</td><td>0.89</td></tr><tr><td>1.3</td><td>0.87</td></tr><tr><td>1.4</td><td>0.85</td></tr><tr><td>1.6</td><td>0.81</td></tr></tbody></table></div> <p>Lastly, we compare again our β 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.</p> <p>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 β. Our dataset basically shows that for high/low Beta, the linear relation between returns and market returns is weaker, compared to medium- β. Which means, that our model is less accurate, when we analyze high/low β Portfolios.</p>	Estimated Betas	Estimated R^2	0.6	0.61	0.7	0.73	0.8	0.83	0.9	0.88	1.0	0.89	1.1	0.88	1.2	0.89	1.3	0.87	1.4	0.85	1.6	0.81	
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1.4	0.85																							
1.6	0.81																							

f)
Build beta-neutral portfolio and plot results vs. market (6 points)

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



g)
Performance comparison of beta-neutral portfolio to market (6 points)

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

<u>Correlation Matrix</u>	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.

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

CAPM regression:

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=====
                        OLS Regression Results
=====
Dep. Variable:                y      R-squared:                0.000
Model:                        OLS      Adj. R-squared:           -0.001
Method:                        Least Squares      F-statistic:           0.007204
Date:                        Mon, 29 Nov 2021      Prob (F-statistic):       0.932
Time:                        22:52:57      Log-Likelihood:         776.27
No. Observations:            696      AIC:                   -1549.
Df Residuals:                694      BIC:                   -1539.
Df Model:                    1
Covariance Type:              nonrobust
=====
                        coef      std err          t      P>|t|      [0.025      0.975]
-----
const                0.0099      0.003      3.260      0.001      0.004      0.016
x1                   0.0057      0.068      0.085      0.932     -0.127      0.139
=====
Omnibus:                35.042      Durbin-Watson:           1.910
Prob(Omnibus):          0.000      Jarque-Bera (JB):        101.569
Skew:                   -0.142      Prob(JB):                8.80e-23
Kurtosis:               4.850      Cond. No.                22.5
=====

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Fama-French 3-factor regression:

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=====
                        OLS Regression Results
=====
Dep. Variable:                y      R-squared:                0.332
Model:                        OLS      Adj. R-squared:           0.329
Method:                        Least Squares      F-statistic:           114.5
Date:                        Mon, 29 Nov 2021      Prob (F-statistic):     3.15e-60
Time:                        22:52:57      Log-Likelihood:         916.58
No. Observations:            696      AIC:                   -1825.
Df Residuals:                692      BIC:                   -1807.
Df Model:                    3
Covariance Type:              nonrobust
=====
                        coef      std err          t      P>|t|      [0.025      0.975]
-----
const                0.0089      0.003      3.566      0.000      0.004      0.014
x1                   0.3886      0.059      6.566      0.000      0.272      0.505
x2                  -1.4095      0.085     -16.581      0.000     -1.576     -1.243
x3                   0.7774      0.087      8.949      0.000      0.607      0.948
=====
Omnibus:                24.117      Durbin-Watson:           2.014
Prob(Omnibus):          0.000      Jarque-Bera (JB):        58.622
Skew:                   0.047      Prob(JB):                1.86e-13
Kurtosis:               4.419      Cond. No.                37.1
=====

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Fama-French 5-factor regression

OLS Regression Results						
=====						
Dep. Variable:	y	R-squared:	0.384			
Model:	OLS	Adj. R-squared:	0.380			
Method:	Least Squares	F-statistic:	86.16			
Date:	Mon, 29 Nov 2021	Prob (F-statistic):	2.42e-70			
Time:	22:52:57	Log-Likelihood:	945.08			
No. Observations:	696	AIC:	-1878.			
Df Residuals:	690	BIC:	-1851.			
Df Model:	5					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.0045	0.002	1.815	0.070	-0.000	0.009
x1	0.5314	0.060	8.800	0.000	0.413	0.650
x2	-1.2419	0.086	-14.419	0.000	-1.411	-1.073
x3	0.3060	0.112	2.724	0.007	0.085	0.527
x4	0.7058	0.119	5.923	0.000	0.472	0.940
x5	1.0264	0.173	5.946	0.000	0.687	1.365
=====						
Omnibus:	6.622	Durbin-Watson:	2.008			
Prob(Omnibus):	0.036	Jarque-Bera (JB):	8.870			
Skew:	-0.057	Prob(JB):	0.0119			
Kurtosis:	3.541	Cond. 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.

i)
Performance analysis of rebalanced portfolio (10 points)

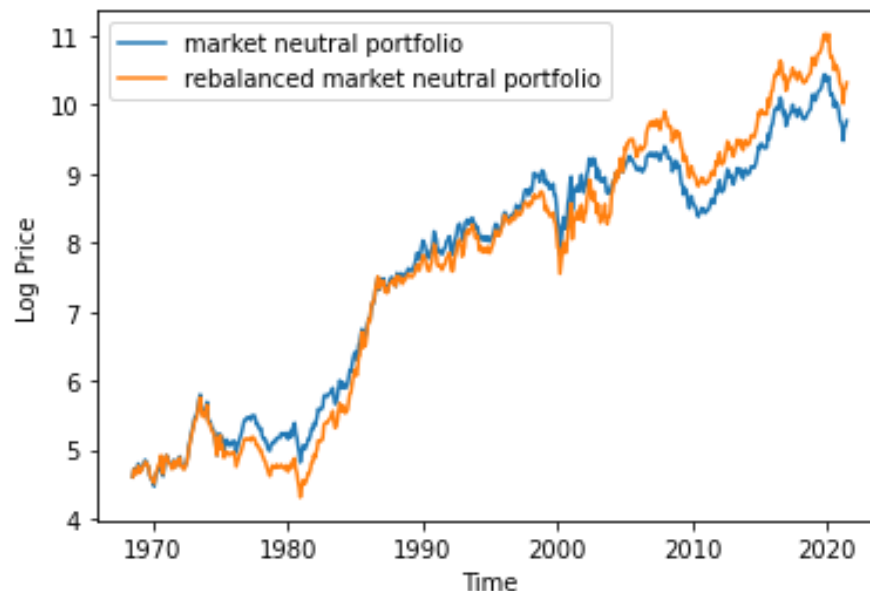
French-Fama 5-factor regression on the monthly rebalanced portfolio:

OLS Regression Results						
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Dep. Variable:	y	R-squared:	0.341			
Model:	OLS	Adj. R-squared:	0.336			
Method:	Least Squares	F-statistic:	65.22			
Date:	Mon, 29 Nov 2021	Prob (F-statistic):	7.29e-55			
Time:	23:51:40	Log-Likelihood:	798.12			
No. Observations:	636	AIC:	-1584.			
Df Residuals:	630	BIC:	-1558.			
Df Model:	5					
Covariance Type:	nonrobust					
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	coef	std err	t	P> t	[0.025	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
x4	0.6319	0.136	4.661	0.000	0.366	0.898
x5	1.1176	0.201	5.550	0.000	0.722	1.513
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Omnibus:	8.412	Durbin-Watson:	1.935			
Prob(Omnibus):	0.015	Jarque-Bera (JB):	11.159			
Skew:	0.125	Prob(JB):	0.00377			
Kurtosis:	3.599	Cond. No.	82.0			
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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)

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

→ $v = 4 \times 1$ 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

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

$$\text{Var}(r_p) = v' * \Sigma * v \quad (\text{iii})$$

→ μ_{risky} , μ_{rf} = mean log returns over 2-year time horizon, Σ = variance-covariance matrix

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

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

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:

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

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

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

Solving for Lambda:

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

(vii) into (iv):

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

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

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

(viii) into (vi):

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

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

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

Time-varying weights of factor model:



based on the optimization solution:

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

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

c)
Report
performance
measures
and run
Fama-
French 5-
factor
regression
(10 points)

Performance measures:

Mean Return: 0.45%

Volatility: 0.86%

Sharpe Ratio: 46.3%

Fama-French 5-factor regression:

OLS Regression Results						
Dep. Variable:	y	R-squared:	0.068			
Model:	OLS	Adj. R-squared:	0.028			
Method:	Least Squares	F-statistic:	1.694			
Date:	Tue, 30 Nov 2021	Prob (F-statistic):	0.142			
Time:	20:32:20	Log-Likelihood:	411.08			
No. Observations:	122	AIC:	-810.2			
Df Residuals:	116	BIC:	-793.3			
Df Model:	5					
Covariance Type:	nonrobust					
	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	Durbin-Watson:	1.533			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	90.741			
Skew:	1.394	Prob(JB):	1.98e-20			
Kurtosis:	6.174	Cond. No.	82.7			

	<p>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.</p>	
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