

Pairs Trading

MGMT 638: future-Driven Investments: Equity

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Overview

- Find stock pairs that usually track together
- When the relationship is broken:
 - Buy the stock that is cheap compared to the usual relationship
 - Short sell the stock that is expensive compared to the usual relationship
- Hope the usual relationship is restored soon.



Model

- $P_1/P_2 \approx \text{constant}$
- When the ratio goes above the constant, it tends to come down.
- When the ratio goes below the constant, it tends to go up.

$$\Delta P_1/P_2 = \begin{cases} + & \text{when } P_1/P_2 < \text{constant} \\ - & \text{when } P_1/P_2 > \text{constant} \end{cases}$$

-Assume the change is larger when the ratio is further from the constant as

$$\Delta P_1/P_2 = k(\text{constant} - P_1/P_2)$$

for a constant $k > 0$.



- The model is equivalent to

$$\Delta P_1/P_2 = a + bP_1/P_2$$

where $a = k \times \text{constant}$, $b = -k$.

- Estimate a and b by linear regression.
- Should get $a > 0$, $b < 0$.
- If so, $\text{constant} = -a/b$.
- Hold asset 1 and short 2 when $P_1/P_2 < -a/b - \text{threshold}$.
- Hold asset 2 and short 1 when $P_1/P_2 > -a/b + \text{threshold}$.

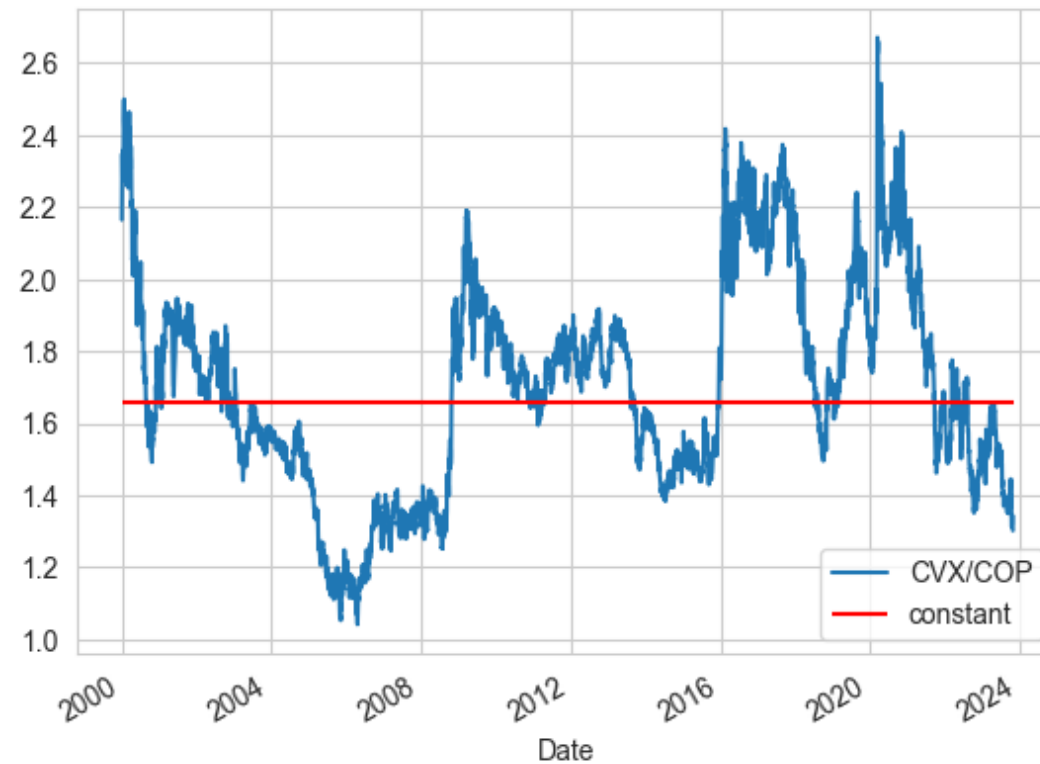


Example

- Chevron (CVX) and Conoco-Phillips (COP) from 2000 on
- Adjusted closing prices from Yahoo Finance
- Compute the price ratio: CVX / COP



```
In [13]: data.ratio.plot(label="CVX/COP")
plt.hlines(
    y=-a/b,
    xmin = data.index[0],
    xmax=data.index[-1],
    color="red",
    label="constant"
)
plt.legend(loc="lower right")
plt.show()
```



Returns

- $-a/b = 1.66$
- Set threshold = 0.2 as an example
- Buy COP and short CVX when CVX / COP is above 1.86
- Buy CVX and short COP when CVX / COP is below 1.46



Market Neutrality

- The pairs strategy is an example of a market neutral strategy, meaning its market beta should be approximately zero.
- If it has a return above the risk-free rate, then adding some of it to the market portfolio can improve performance relative to holding the market.
- This is the same as saying that the strategy has a positive alpha.
- It is also the same as saying

$$\text{Sharpe ratio of strategy} > \text{Sharpe ratio of market} \\ \times \text{correlation with market}$$

- Get the market return from Ken French's data library.




```
In [15]: print(f"mean return of pairs strategy = {252*data.ret.mean():.2%} annualized"  
print(f"correlation of pairs strategy with market = {data.ret.corr(data.mkt):
```

```
mean return of pairs strategy = 5.41% annualized  
correlation of pairs strategy with market = 4.75%
```



Avoid Look-Ahead Bias

- Compute the parameter of the strategy (the constant $-a/b$) from data through 2015
- Test the strategy from 2015 on.



```
In [18]: print(f"mean return of pairs strategy = {252*future.ret.mean():.2%} annualized")
print(f"correlation of pairs strategy with market = {future.ret.corr(future.m
```

```
mean return of pairs strategy = 6.15% annualized
correlation of pairs strategy with market = 10.74%
```



Alpha and Beta

- $\text{beta} = \text{corr with market excess return} \times \text{std dev of strategy} / \text{std dev of market}$
- $\text{alpha} = \text{mean return} - \text{beta} * \text{mean market excess return}$



```
In [19]: beta = future.ret.corr(future.mkt) * future.ret.std() / future.mkt.std()
alpha = future.ret.mean() - beta * future.mkt.mean()

print(f"beta is {beta:.4f}")
print(f"annualized alpha is {252*alpha:.2%}")
```

```
beta is 0.0011
annualized alpha is 4.87%
```



Regressions in python

- use statsmodels.formula.api
- `smf.ols("model", data).fit().summary()`



```
In [20]: import statsmodels.formula.api as smf
smf.ols("ret~mkt", future).fit().summary()
```

Out[20]:

OLS Regression Results						
Dep. Variable:		ret		R-squared:		0.012
Model:		OLS		Adj. R-squared:		0.011
Method:		Least Squares		F-statistic:		25.15
Date:		Wed, 25 Oct 2023		Prob (F-statistic):		5.75e-07
Time:		12:27:27		Log-Likelihood:		6431.3
No. Observations:		2157		AIC:		-1.286e+04
Df Residuals:		2155		BIC:		-1.285e+04
Df Model:		1				
Covariance Type:		nonrobust				
	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.0002	0.000	0.758	0.448	-0.000	0.001
mkt	0.0011	0.000	5.015	0.000	0.001	0.002
Omnibus:		438.977	Durbin-Watson:		1.951	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		11512.699	
Skew:		0.267	Prob(JB):		0.00	
Kurtosis:		14.305	Cond. No.		1.20	

