

Final Project - Forecasting Gold Price

Aliaksandr Panko

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1 Objectives

To forecast gold price evolution we would like to compare a linear regression estimated by the Ordinary Least Squares (OLS) method and another model.

2 Data

Gold prices are given:

- [Data Source](#)

- Frequency: Annual data
- Period: 1978 - 2014

3 Linar Regression

In the first part of the project OLS linear regression is used.

1. Explanatory variable: Previous gold price
2. Explained variable: Current gold price

The regression equation is:

$$gold_t = \alpha + \beta gold_{t-1}$$

3.1 Results of the regression

Original:

OLS Regression Results						
Dep. Variable:	gold	R-squared:	0.906			
Model:	OLS	Adj. R-squared:	0.903			
Method:	Least Squares	F-statistic:	326.3			
Date:	Tue, 20 Feb 2018	Prob (F-statistic):	5.31e-19			
Time:	22:31:23	Log-Likelihood:	-222.73			
No. Observations:	36	AIC:	449.5			
Df Residuals:	34	BIC:	452.6			
Df Model:	1					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
const	36.1122	35.585	1.015	0.317	-36.205	108.429
gold_lag	0.9882	0.055	18.064	0.000	0.877	1.099
Omnibus:	5.980	Durbin-Watson:	1.175			
Prob(Omnibus):	0.050	Jarque-Bera (JB):	4.617			
Skew:	0.654	Prob(JB):	0.0994			
Kurtosis:	4.169	Cond. No.	1.15e+03			

Logarithm:

OLS Regression Results						
Dep. Variable:	log_gold	R-squared:	0.869			
Model:	OLS	Adj. R-squared:	0.865			
Method:	Least Squares	F-statistic:	225.0			
Date:	Tue, 20 Feb 2018	Prob (F-statistic):	1.48e-16			
Time:	22:32:31	Log-Likelihood:	8.9141			
No. Observations:	36	AIC:	-13.83			
Df Residuals:	34	BIC:	-10.66			
Df Model:	1					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
const	0.4074	0.386	1.056	0.299	-0.377	1.192
log_gold_lag	0.9420	0.063	15.001	0.000	0.814	1.070
Omnibus:	10.138	Durbin-Watson:	1.234			
Prob(Omnibus):	0.006	Jarque-Bera (JB):	9.400			
Skew:	0.993	Prob(JB):	0.00910			
Kurtosis:	4.523	Cond. No.	75.1			

As we can see in both models constant is not significant at 95% significance level (p_value > 0.05)

Based on R^2 we can conclude that in both cases the model describes gold prices evaluation rather good(gold price in the previous period explains 90.56 % and 86.87 % of today's evolution respectively)

So, it is a good idea to implement this model without intercept.

3.2 Model Without Alpha

The regression equation is:

$$gold_t = \beta gold_{t-1}$$

This model gives us next result:

OLS Regression Results						
Dep. Variable:	gold	R-squared:	0.969			
Model:	OLS	Adj. R-squared:	0.969			
Method:	Least Squares	F-statistic:	1110.			
Date:	Tue, 20 Feb 2018	Prob (F-statistic):	4.23e-28			
Time:	22:45:16	Log-Likelihood:	-223.26			
No. Observations:	36	AIC:	448.5			
Df Residuals:	35	BIC:	450.1			
Df Model:	1					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
gold_lag	1.0340	0.031	33.313	0.000	0.971	1.097
Omnibus:	3.876	Durbin-Watson:	1.193			
Prob(Omnibus):	0.144	Jarque-Bera (JB):	3.211			
Skew:	0.182	Prob(JB):	0.201			
Kurtosis:	4.417	Cond. No.	1.00			

The result is extremely good, 97% of gold price evaluation can be explained by this model! However Durbin-Watson test and Jarque-Bera test signal that there is autocorrelation problem and residuals are not normally distributed.

Actually, it is well-known problem: "since stock prices tend not to change too radically from one day to another, the prices from one day to the next could potentially be highly correlated, even though there is little useful information in this observation. In order to avoid autocorrelation issues, the easiest solution in finance is to simply convert a series of historical prices into a series of percentage-price changes from day to day. " [ref. Investopedia](#)

In a conclusion, I would rather use model with percentage that are mentioned above.

3.3 Multiple regression

I decided to take into consideration next factors:

1. Producer Price Index for All Commodities [[PPIACO](#)]
2. Wilshire 5000 Total Market Index [[WILL5000IND](#)]

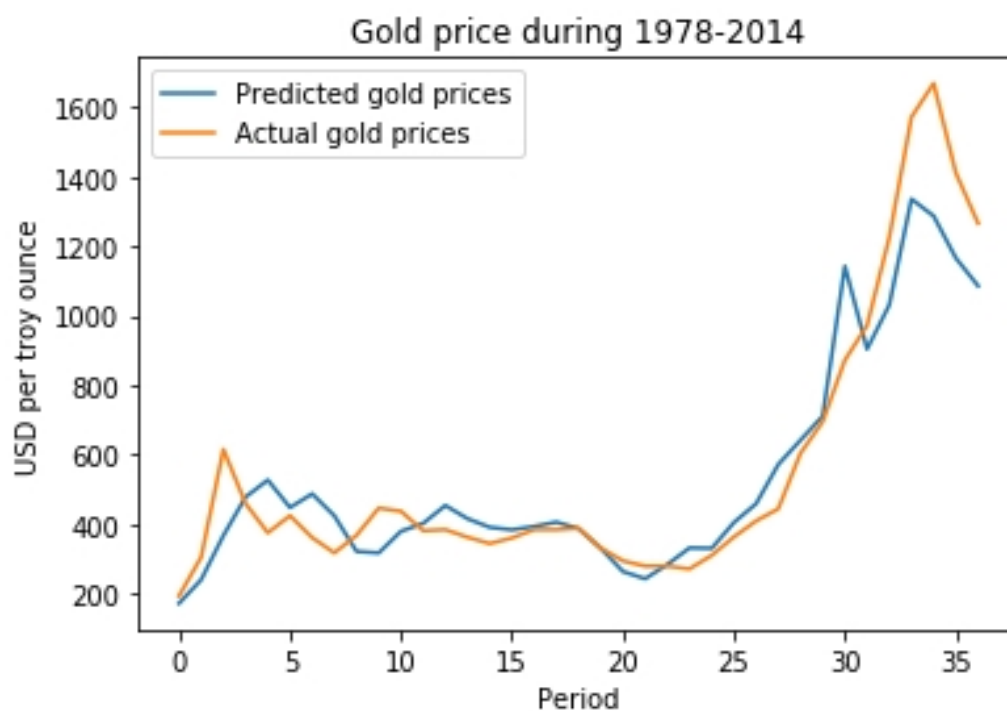
Frankly speaking, it was not so easy to find out these factors. I also tried money supply and interest rates but they were not significant.

Likely in the end the model gives rather good result:

OLS Regression Results						
Dep. Variable:	y	R-squared:	0.878			
Model:	OLS	Adj. R-squared:	0.871			
Method:	Least Squares	F-statistic:	122.6			
Date:	Wed, 21 Feb 2018	Prob (F-statistic):	2.83e-16			
Time:	00:10:59	Log-Likelihood:	9.5962			
No. Observations:	37	AIC:	-13.19			
Df Residuals:	34	BIC:	-8.360			
Df Model:	2					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
const	-11.5109	1.293	-8.905	0.000	-14.138	-8.884
PPIACO	3.9755	0.301	13.208	0.000	3.364	4.587
WILL5000IND	-0.6026	0.069	-8.779	0.000	-0.742	-0.463
Omnibus:	1.471	Durbin-Watson:	1.014			
Prob(Omnibus):	0.479	Jarque-Bera (JB):	0.910			
Skew:	0.382	Prob(JB):	0.634			
Kurtosis:	3.073	Cond. No.	237.			

As we can see all factors are significant at 1% level and 87.8% data are explained by the model. Nevertheless Durbin-Watson test still is not perfect that means autocorrelation existence.

Let us compare the results visually:



As we can see it is good enough.

4 ARMA model

Since we have rather good prediction taking into account only previous value I decided to try ARMA model model for predicting gold price evolution.

4.1 Stationarity

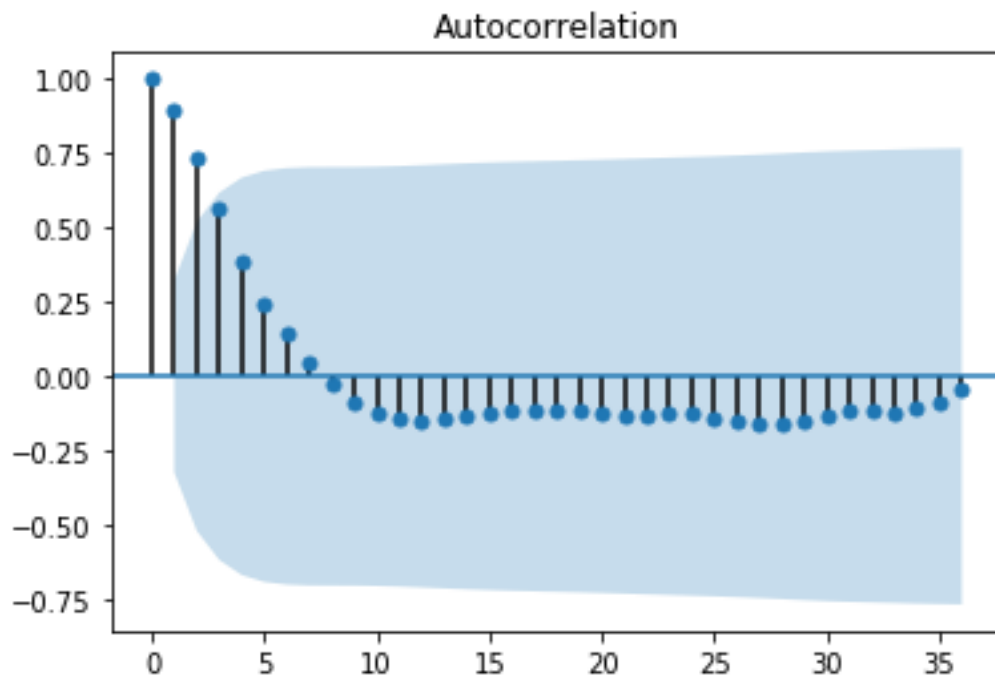
Firstly, we need to check stationarity. For this purposes I used Augmented Dickey-Fuller unit root test.

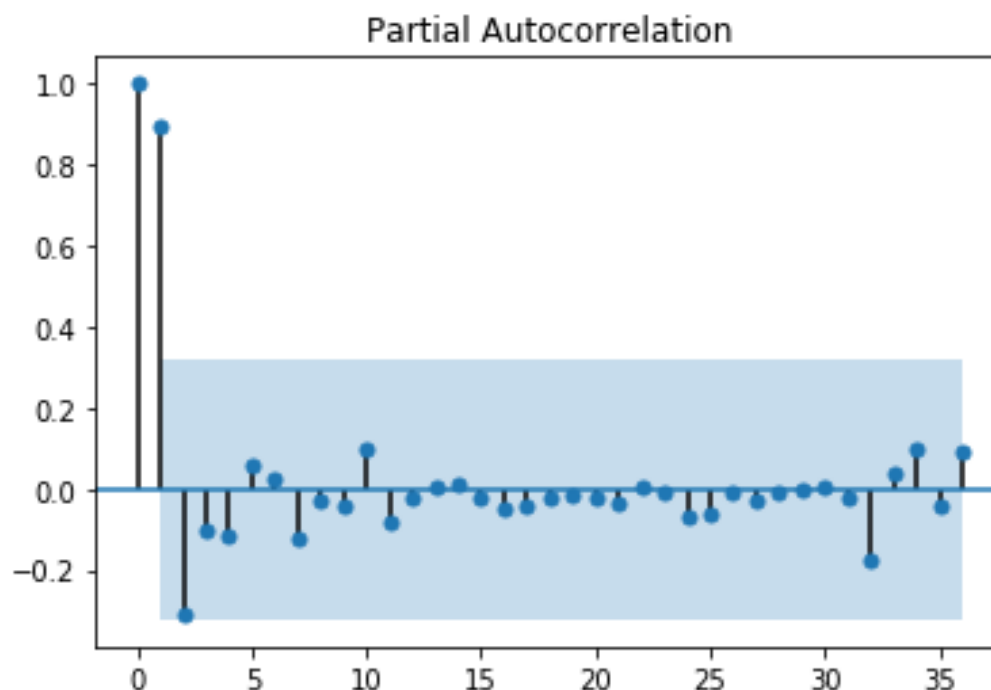
The null hypothesis of the **Augmented Dickey-Fuller** is that there is a unit root, with the alternative that there is no unit root.

p.value is 3.13184167501e-06 - this means that we reject H0 and assume that the **process is stationary**.

4.2 ACF & PACF

Secondly, we need to check ACF and PACF and use Box-Jenkins method.

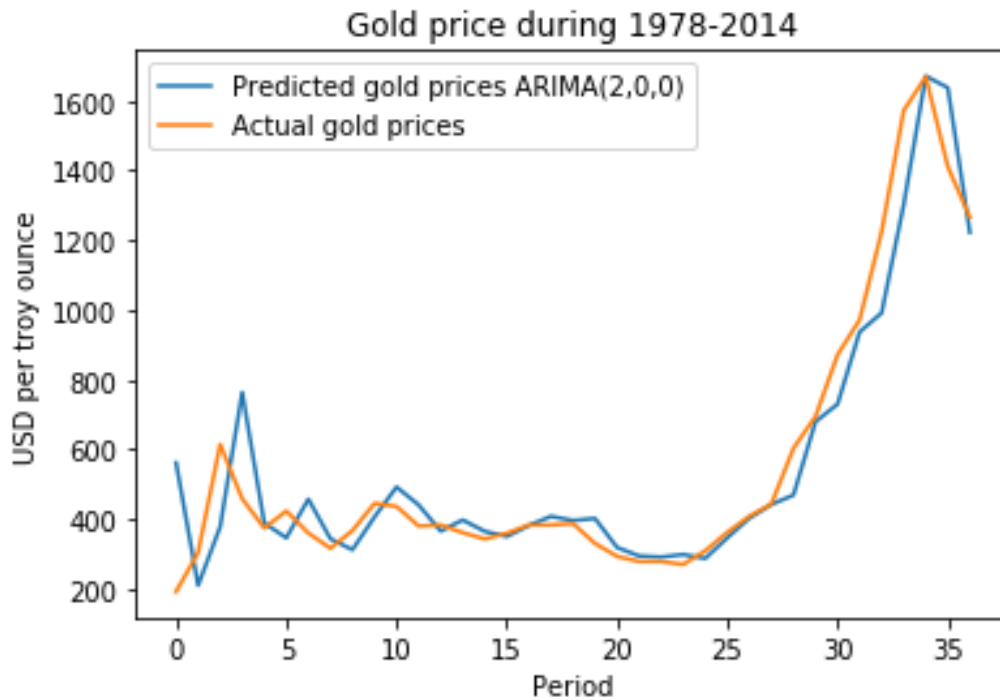




It's clear that ACF decays exponentially and PACF has 2 significant spikes. Since differencing is not required, the parameters of the model are **ARIMA(2,0,0)**.

Let us have a look at the results:

ARMA Model Results						
Dep. Variable:	y	No. Observations:	37			
Model:	ARMA(2, 0)	Log Likelihood	-226.846			
Method:	css-mle	S.D. of innovations	106.968			
Date:	Wed, 21 Feb 2018	AIC	461.692			
Time:	01:02:23	BIC	468.136			
Sample:	0	HQIC	463.964			
	coef	std err	z	P> z	[0.025	0.975]
const	562.3753	214.407	2.623	0.013	142.145	982.606
ar.L1.y	1.4218	0.146	9.712	0.000	1.135	1.709
ar.L2.y	-0.4938	0.157	-3.149	0.003	-0.801	-0.186
Roots						
	Real	Imaginary	Modulus	Frequency		
AR.1	1.2213	+0.0000j	1.2213	0.0000		
AR.2	1.6584	+0.0000j	1.6584	0.0000		



As we can see the model describes the actual values rather good.

5 Forecast

5.1 Multiple regression

$$current_price = \exp(-11.5109 + 3.9755 * \log(ppiaco) - 0.6026 * \log(will5000ind))$$

Assuming that in 2015 $ppiaco$ was 190.4 and market index ($will5000ind$) was 90.66, we can find gold price in 2015 = 766 USD per troy ounce.

5.2 ARIMA

The designed model gives us price value **1143.79641724** for the next year.

I must say that the difference between two models is considerable. But as R^2 says ARIMA result is a little bit more reliable.