

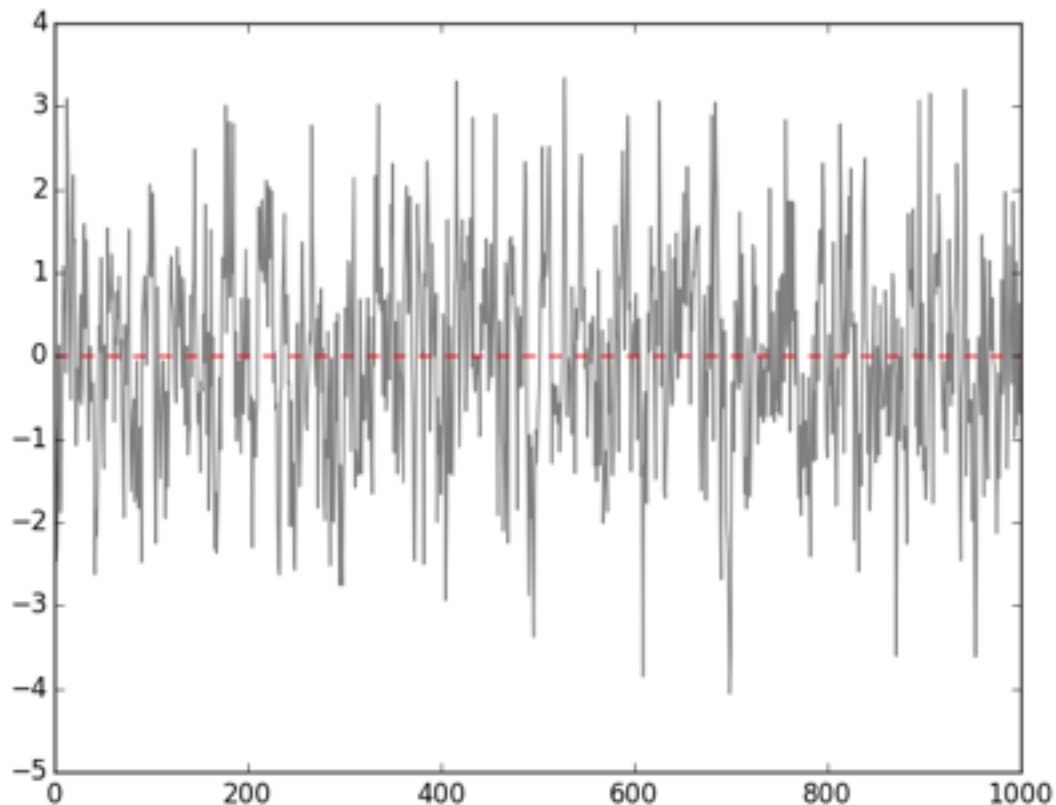
Problem1:

a.p=0.5 is included in 95% confidence interval

the statistic summary for p=0.5

OLS Regression Results						
=====						
Dep. Variable:	y		R-squared:	0.261		
Model:	OLS		Adj. R-squared:	0.261		
Method:	Least Squares		F-statistic:	352.8		
Date:	Sun, 02 Nov 2014		Prob (F-statistic):	1.27e-67		
Time:	17:05:10		Log-Likelihood:	-1434.0		
No. Observations:	999		AIC:	2872.		
Df Residuals:	997		BIC:	2882.		
Df Model:		1				
=====						
	coef	std err	t	P> t	[95.0% Conf. Int.]	

const	0.0259	0.032	0.805	0.421	-0.037	0.089
x1	0.5114	0.027	18.783	0.000	0.458	0.565
=====						
Omnibus:	0.941		Durbin-Watson:	2.010		
Prob(Omnibus):	0.625		Jarque-Bera (JB):	0.989		
Skew:	-0.017		Prob(JB):	0.610		
Kurtosis:	2.849		Cond. No.	1.19		



Problem1:

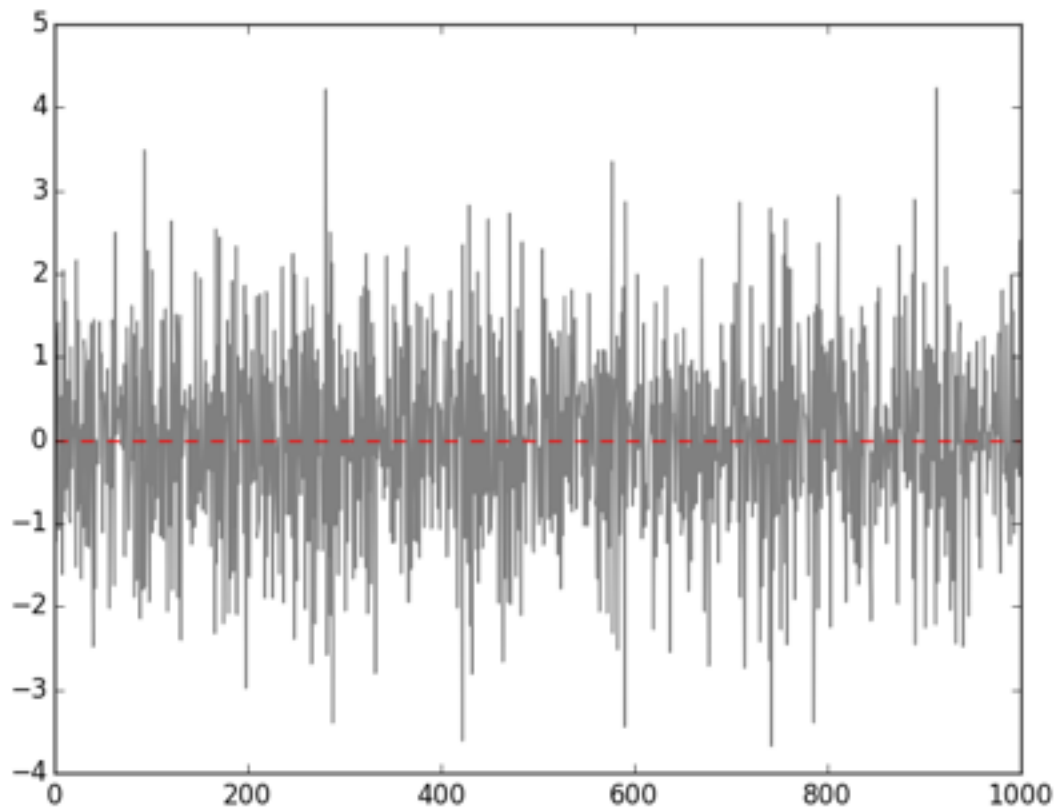
b.p=-0.5 is included in 95% confidence interval

the statistic summary for p=-0.5

OLS Regression Results

=====						
Dep. Variable:	y	R-squared:	0.273			
Model:	OLS	Adj. R-squared:	0.272			
Method:	Least Squares	F-statistic:	374.8			
Date:	Sun, 02 Nov 2014	Prob (F-statistic):	3.98e-71			
Time:	17:09:56	Log-Likelihood:	-1399.5			
No. Observations:	999	AIC:	2803.			
Df Residuals:	997	BIC:	2813.			
Df Model:		1				
=====						
	coef	std err	t	P> t	[95.0% Conf. Int.]	

const	0.0370	0.031	1.191	0.234	-0.024	0.098
x1	-0.5227	0.027	-19.359	0.000	-0.576	-0.470
=====						
Omnibus:	0.183	Durbin-Watson:	2.015			
Prob(Omnibus):	0.912	Jarque-Bera (JB):	0.134			
Skew:	0.026	Prob(JB):	0.935			
Kurtosis:	3.024	Cond. No.	1.15			
=====						



Problem1:

c.p=1 is included in 95% confidence interval

the statistic summary for p=1

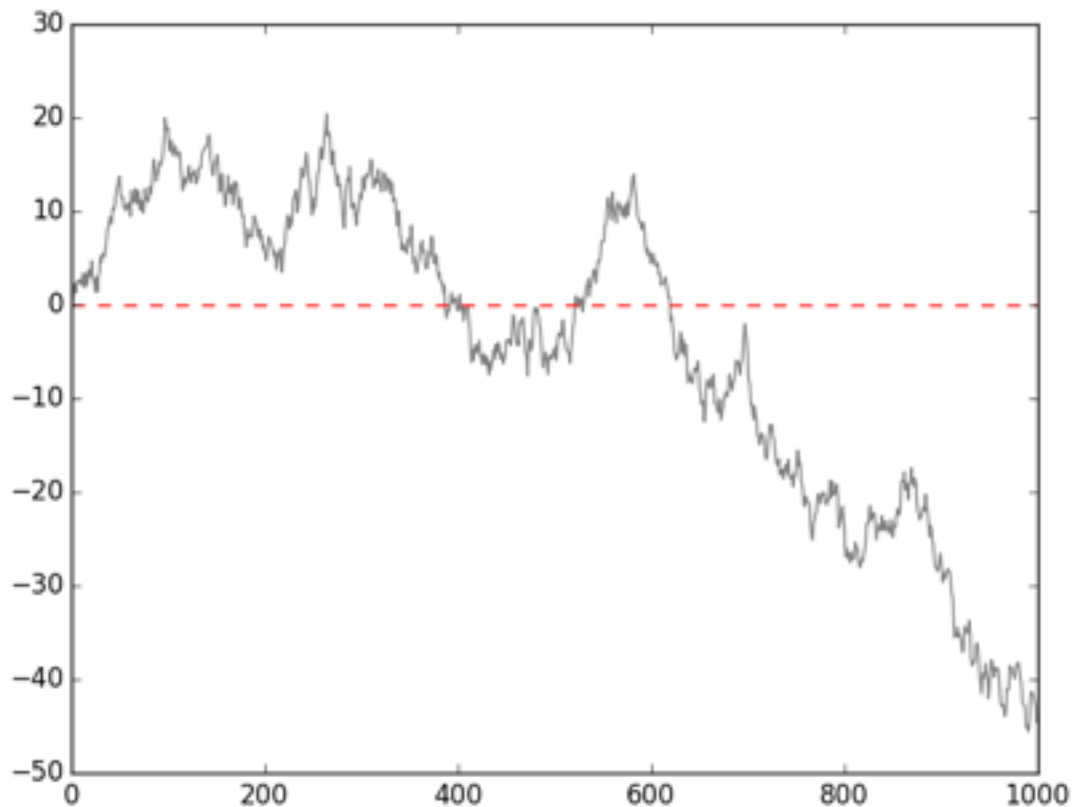
OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.985
Model:                  OLS    Adj. R-squared:       0.985
Method:                 Least Squares    F-statistic:      6.745e+04
Date:                   Sun, 02 Nov 2014    Prob (F-statistic):    0.00
Time:                   17:11:43    Log-Likelihood:      -1428.9
No. Observations:       999    AIC:                2862.
Df Residuals:           997    BIC:                2872.
=====
```

Df Model: 1

```
=====
              coef      std err          t      P>|t|      [95.0% Conf. Int.]
-----
const         0.0797      0.047        1.690      0.091      -0.013      0.172
x1            0.9922      0.004       259.712      0.000      0.985      1.000
=====
```

```
=====
Omnibus:          5.150    Durbin-Watson:          2.039
Prob(Omnibus):    0.076    Jarque-Bera (JB):        3.979
Skew:             0.009    Prob(JB):                0.137
Kurtosis:         2.691    Cond. No.:               18.2
=====
```

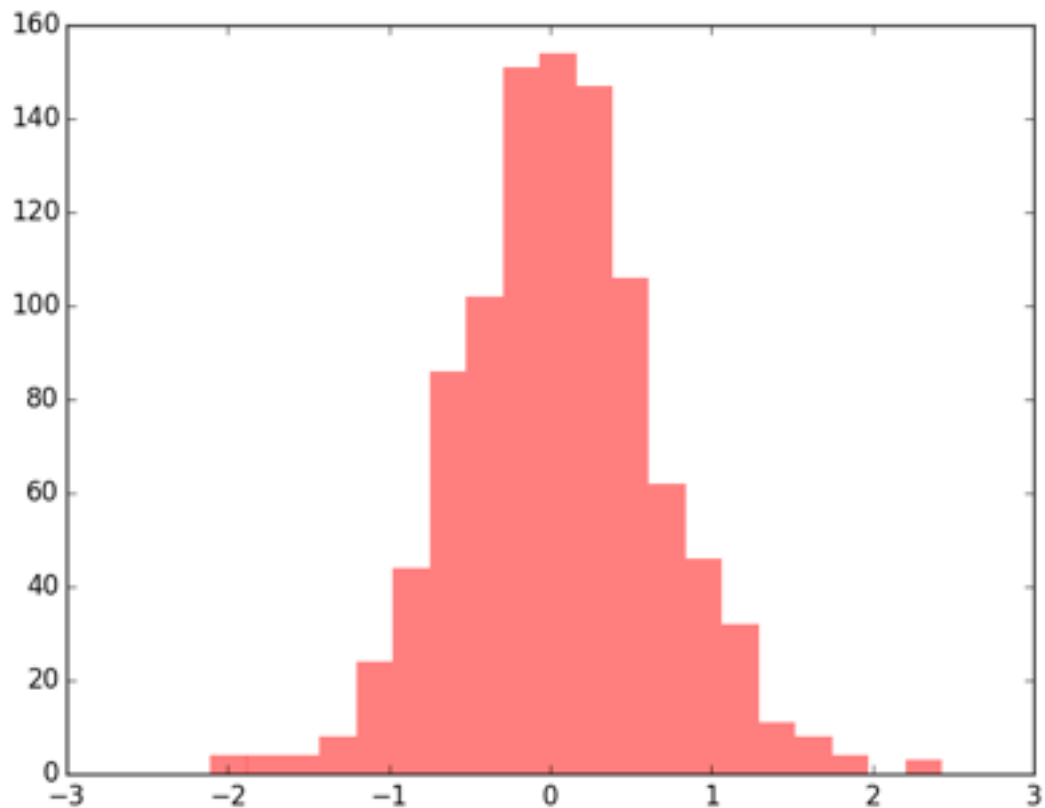


Problem2:**a. regression for Walk2~Walk1****answer for question a**

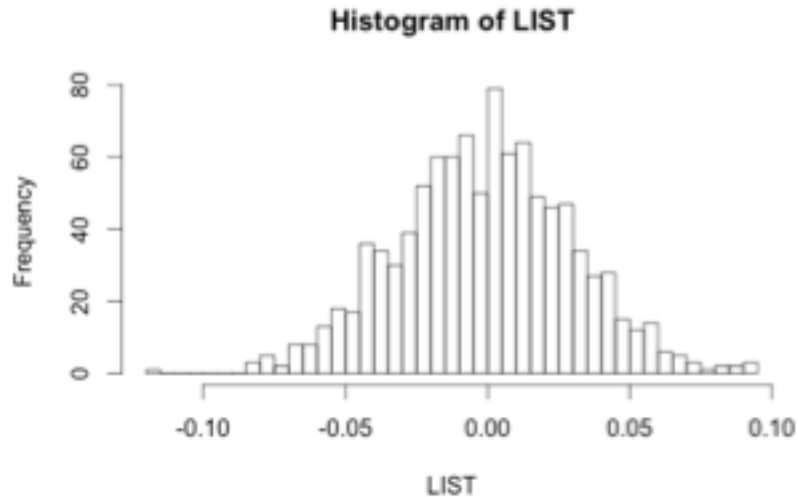
OLS Regression Results

=====					
Dep. Variable:	y	R-squared:	0.376		
Model:	OLS	Adj. R-squared:	0.375		
Method:	Least Squares	F-statistic:	600.9		
Date:	Sun, 02 Nov 2014	Prob (F-statistic):	2.95e-104		
Time:	17:13:01	Log-Likelihood:	-3394.6		
No. Observations:	1000	AIC:	6793.		
Df Residuals:	998	BIC:	6803.		
	Df Model:		1		
=====					
	coef	std err	t	P> t	[95.0% Conf. Int.]

const	-16.7224	0.326	-51.220	0.000	-17.363 -16.082
x1	-0.6585	0.027	-24.514	0.000	-0.711 -0.606
=====					
Omnibus:	30.400	Durbin-Watson:	0.026		
Prob(Omnibus):	0.000	Jarque-Bera (JB):	32.476		
Skew:	0.439	Prob(JB):	8.87e-08		
Kurtosis:	3.096	Cond. No.	17.4		
=====					

b. histogram for coefficient**answer for question b**

comparing to the 3.c in HW4, the histogram in question B has the same center(0) and about the same shape, but more steep (smaller standard deviation)



c. histogram for coefficient

I would say yes, since two variables are independent but the R square are quite big.

Actually, non-stationary time series sometimes could have R square be very close to 1. In this case, I think there could be independent unit roots and there's no correlation but has significant statistical number.