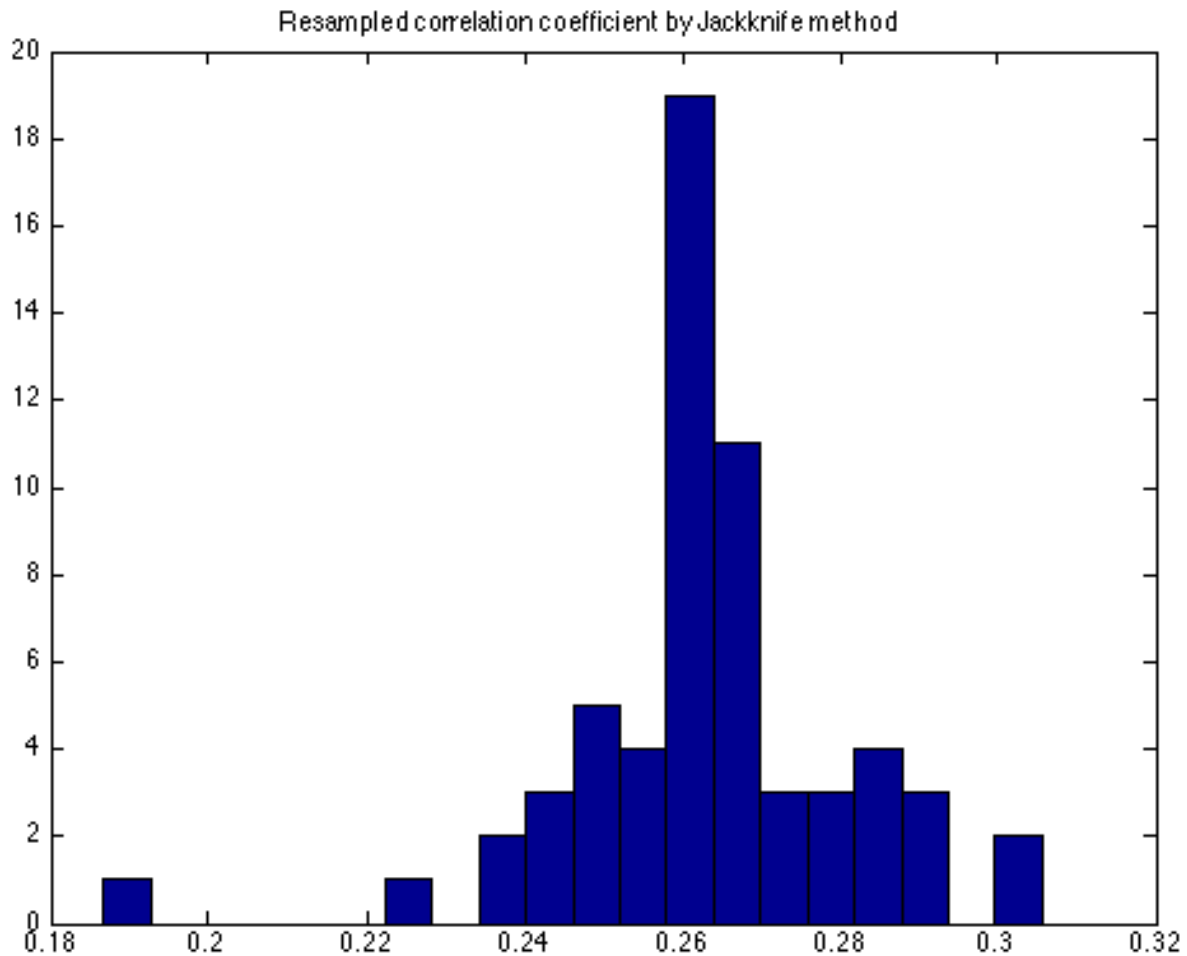


Problem 1. Uncertainties of statistical parameters

#1: Jackknife method for correlation coefficient

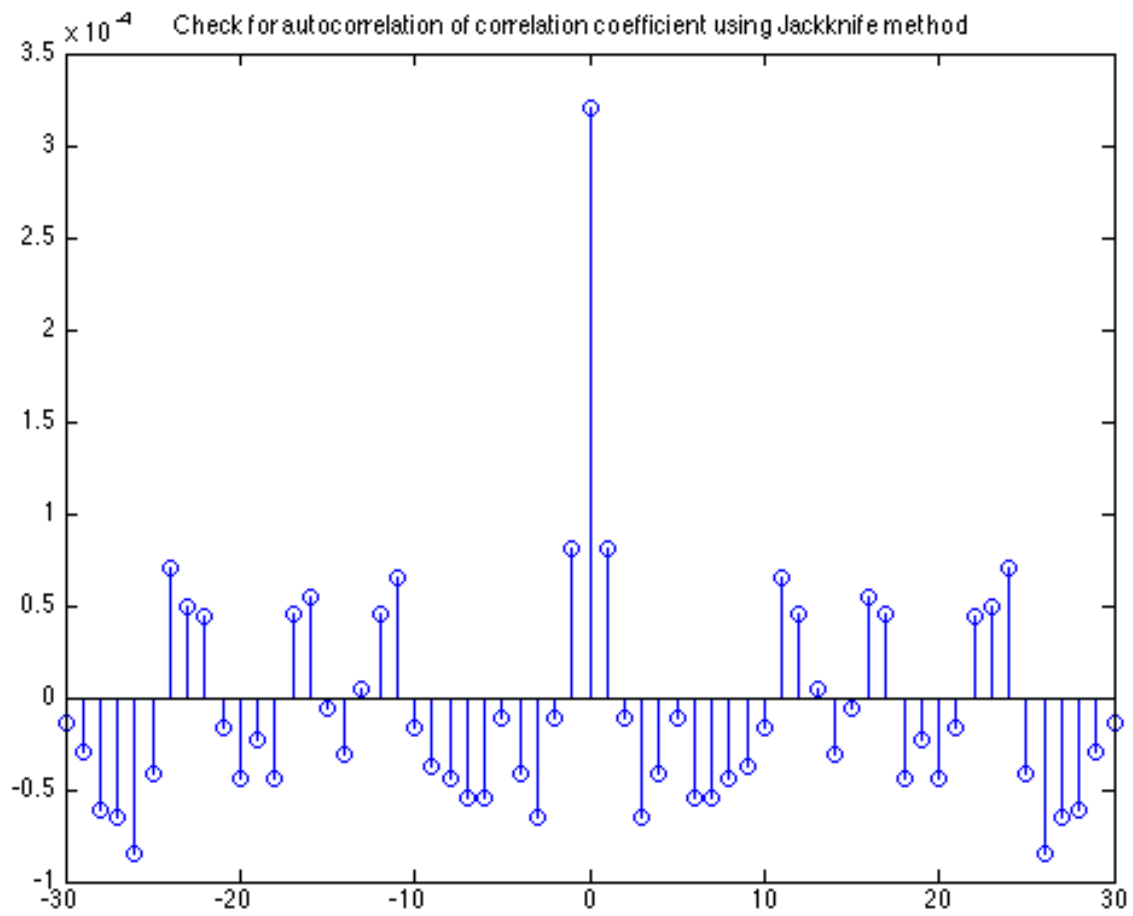
Mean: 0.2631

Standard deviation: 0.0180



95% confidence interval of resampled parameter: [0.2255, 0.3019]

The Jackknife method assumes that observations are independent of each other. We can check if this is true by performing an autocorrelation analysis.

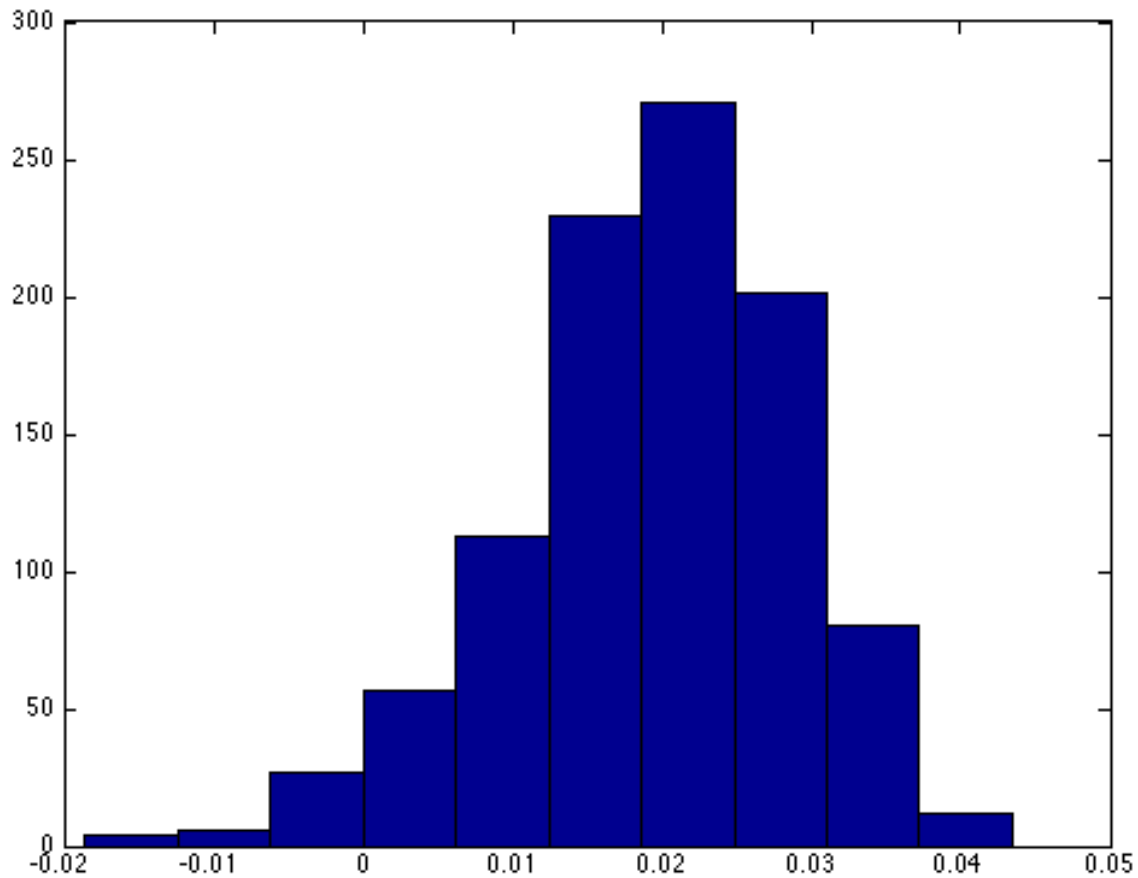


From this plot (lag = 30 days), the correlation coefficient does seem to be autocorrelated because the data is symmetrical and not random.

#2: Bootstrap method for least-squares regression slope

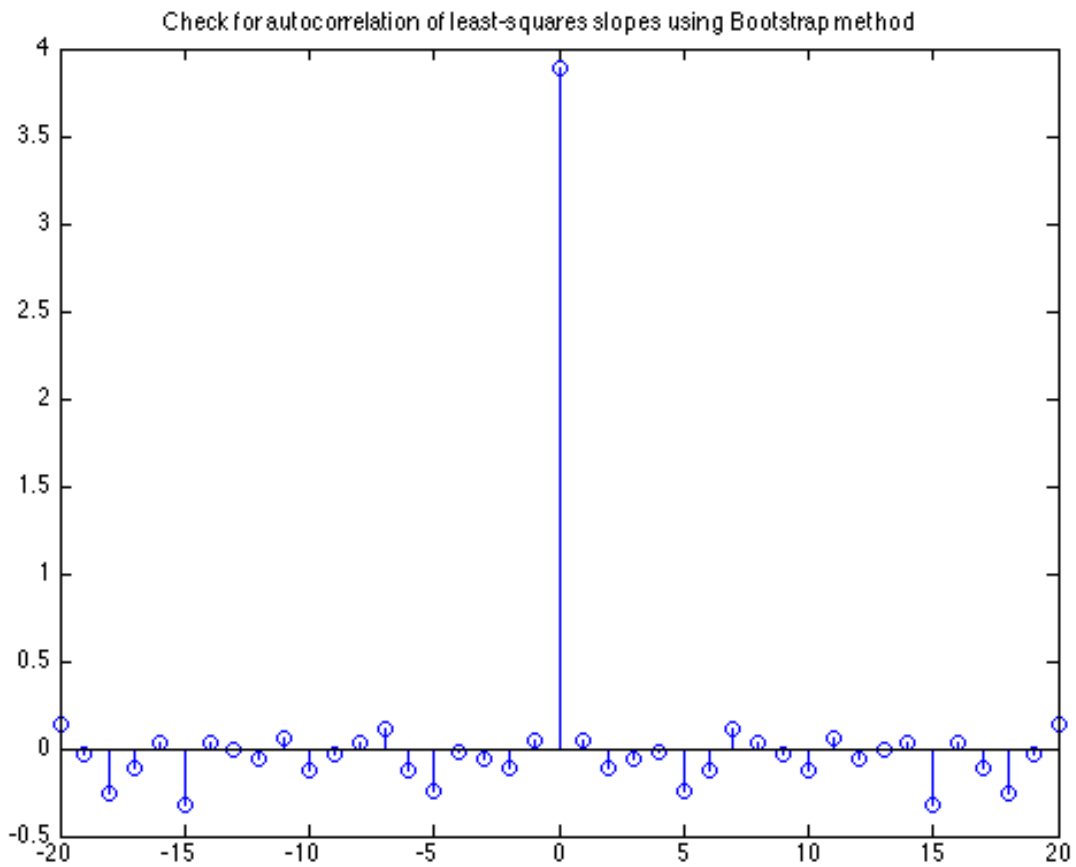
Mean: 0.0192

Standard deviation: 0.0095



95% confidence interval of resampled parameter: [-0.0022 0.0357]

The Bootstrap method assumes that observations are independent of each other. We can check if this is true by performing an autocorrelation analysis.

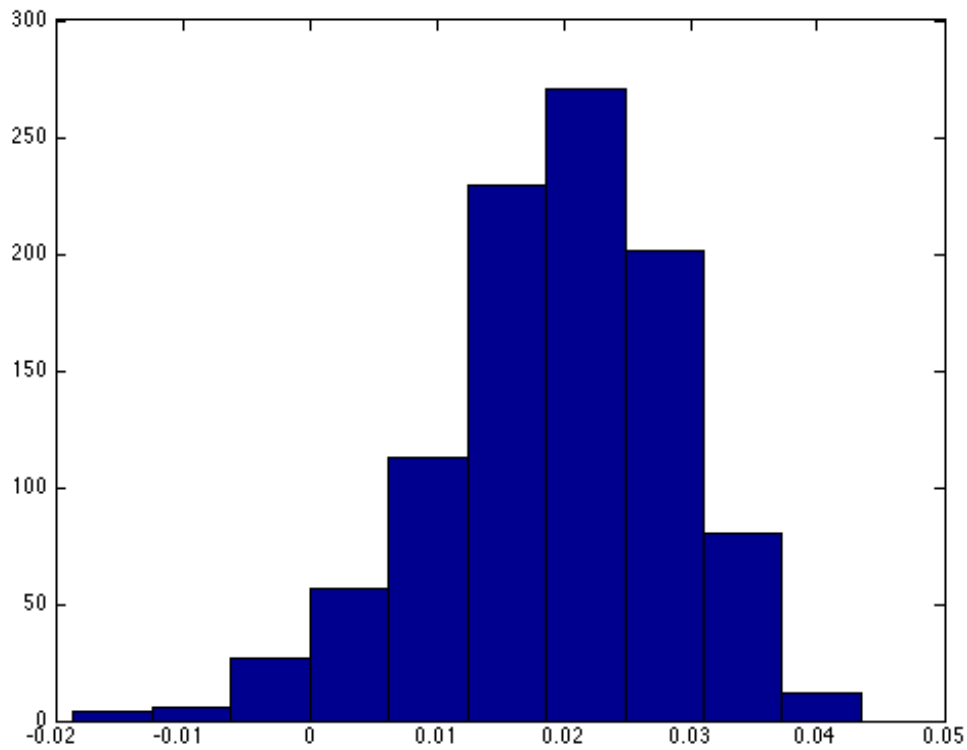


From this plot (lag = 30 days), the slope values do seem to be auto-correlated because the residuals are symmetrical and not random.

#3: Jackknife method for least-squares regression slope

Mean: 3.3701

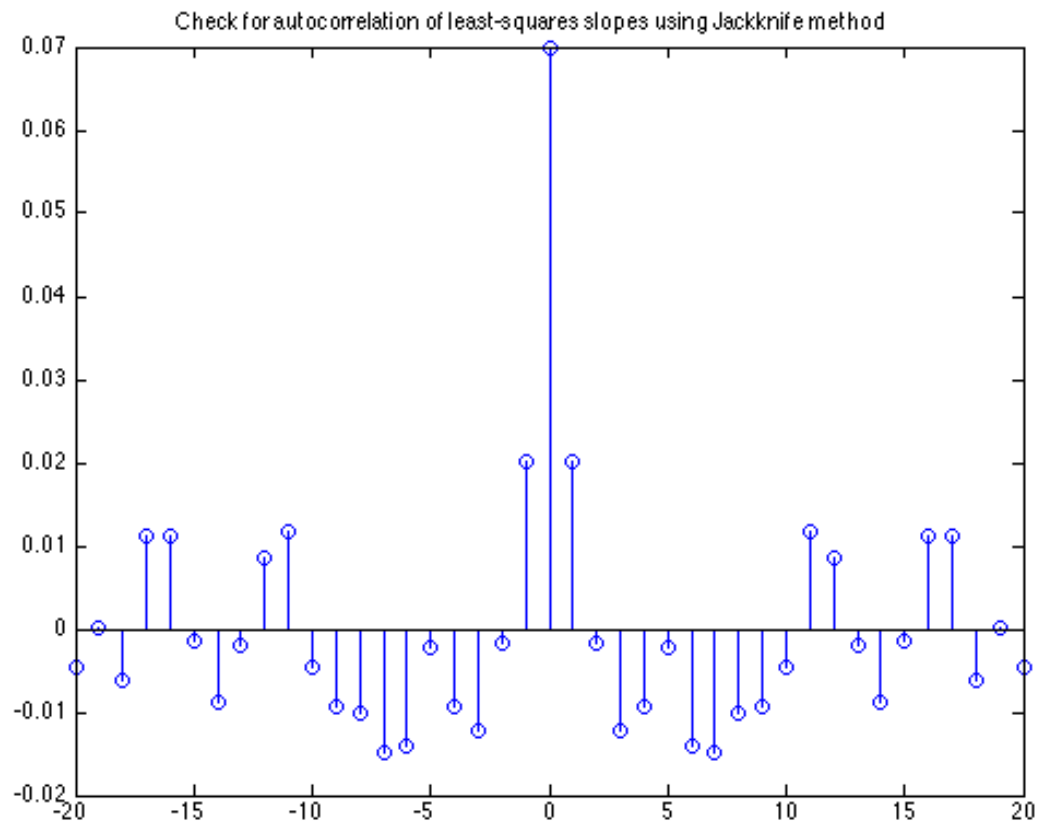
Standard deviation: 0.2664



95% confidence interval of resampled parameter: [0.0184, 0.0233]

The Jackknife method assumes that observations are independent of each other. We can check if this is true by performing an autocorrelation analysis.

Autocorrelation graph (same as #1)

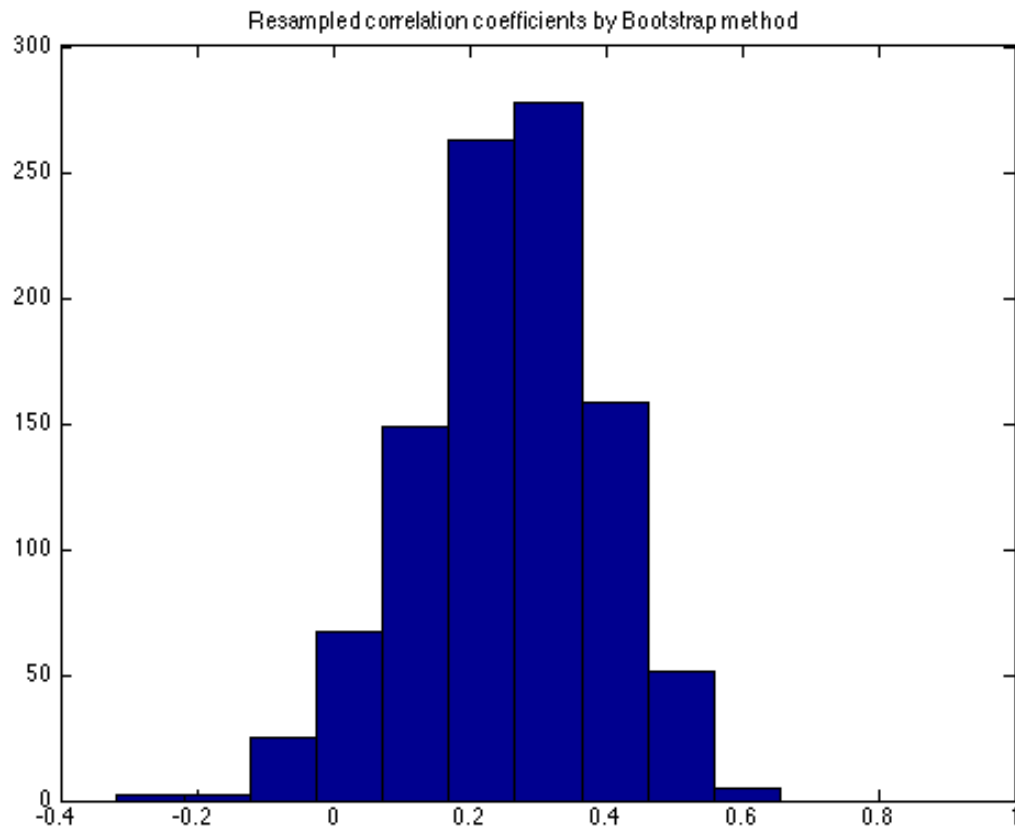


From this plot (lag = 30 days), the slope values do seem to be auto-correlated because the residuals are symmetrical and not random.

#4: Bootstrap method for correlation coefficient

Mean: 0.2600

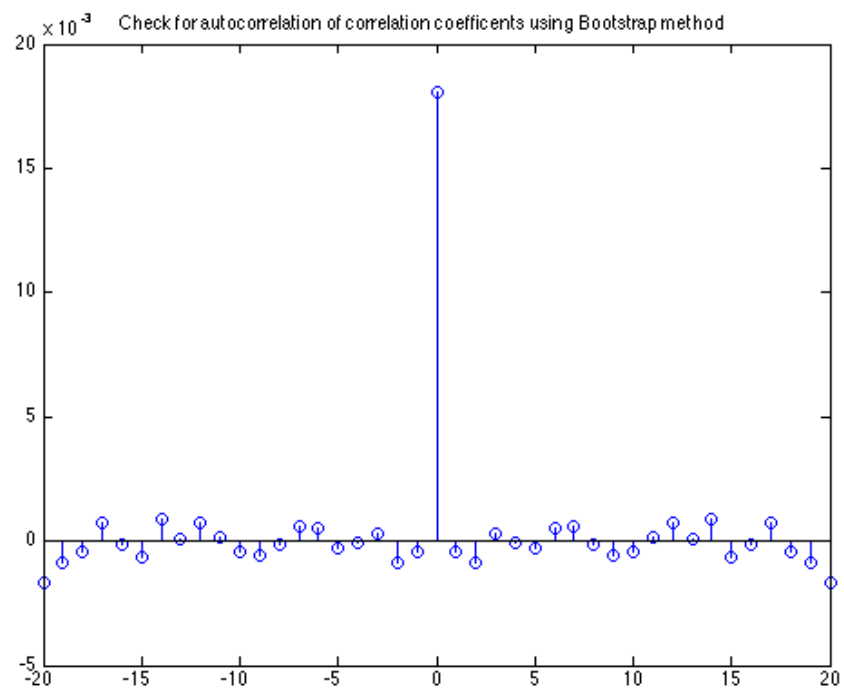
Standard deviation: 0.1321



95% confidence interval of resampled parameter: [-0.0206, 0.5217]

The Bootstrap method assumes that observations are independent of each other. We can check if this is true by performing an autocorrelation analysis.

Autocorrelation graph (same as #2)



From this plot (lag = 30 days), the correlation coefficient values do seem to be auto-correlated because the residuals are symmetrical and not random.

Problem 2. Fitting errors – Is a 1st or 3rd order polynomial fit better for the data?

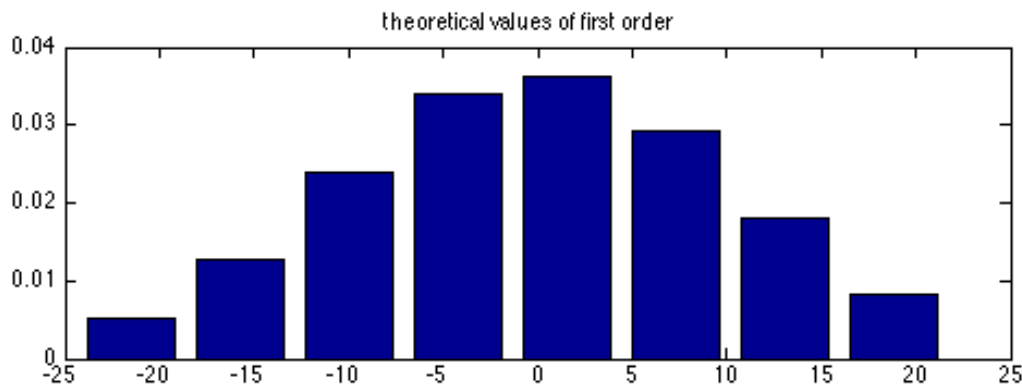
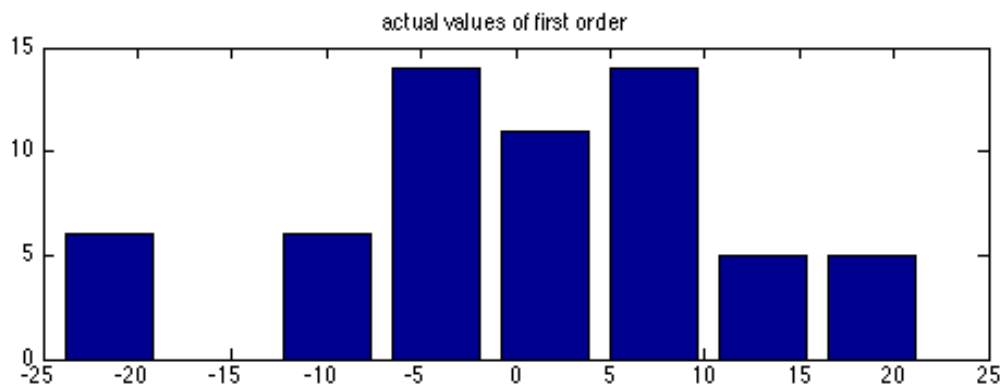
#1: First-order polynomial fit

Checking if residuals are normally distributed:

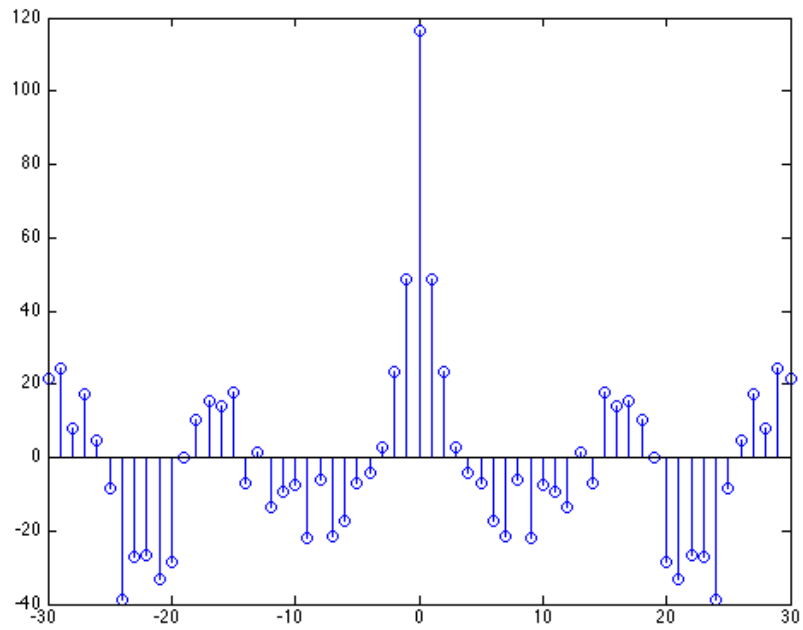
Calculated chi-squared value is 17.8021; critical chi-squared value is 11.0705

Calculated value < critical value means the null hypothesis cannot be rejected

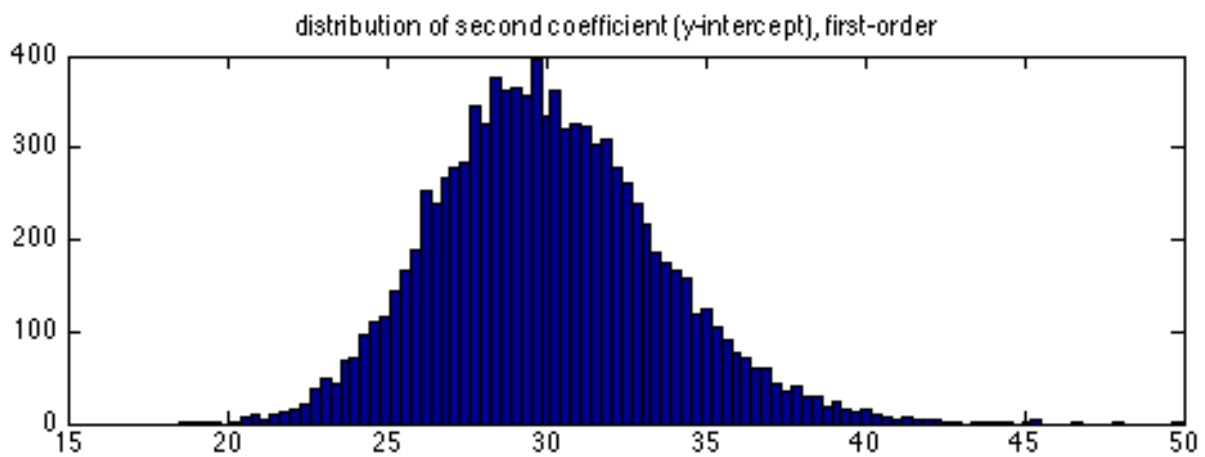
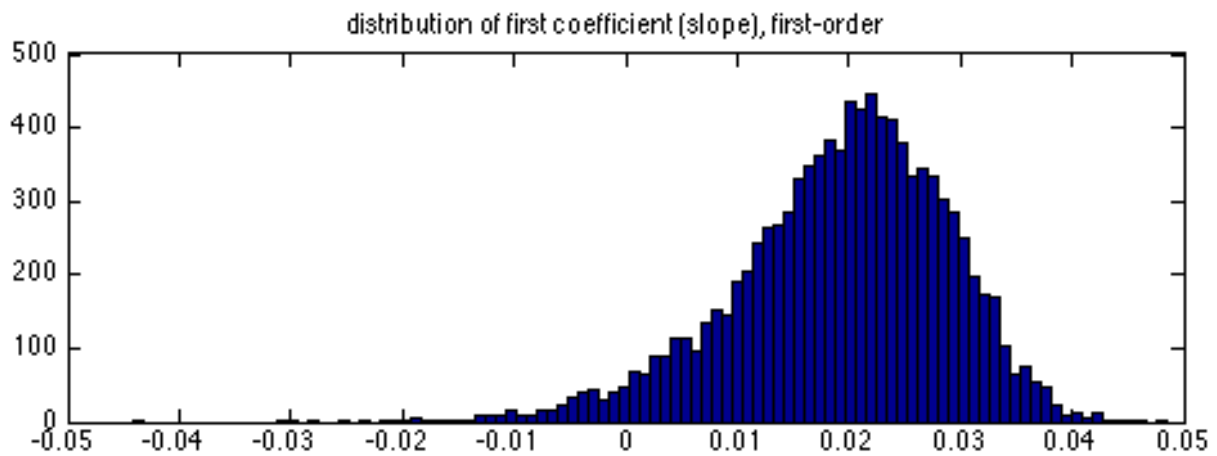
Thus, residuals are normally distributed



Autocorrelation analysis of residuals:



From this plot (lag = 30 days), the residuals for the first-order fit do seem to be auto-correlated because the results are symmetrical and not random.



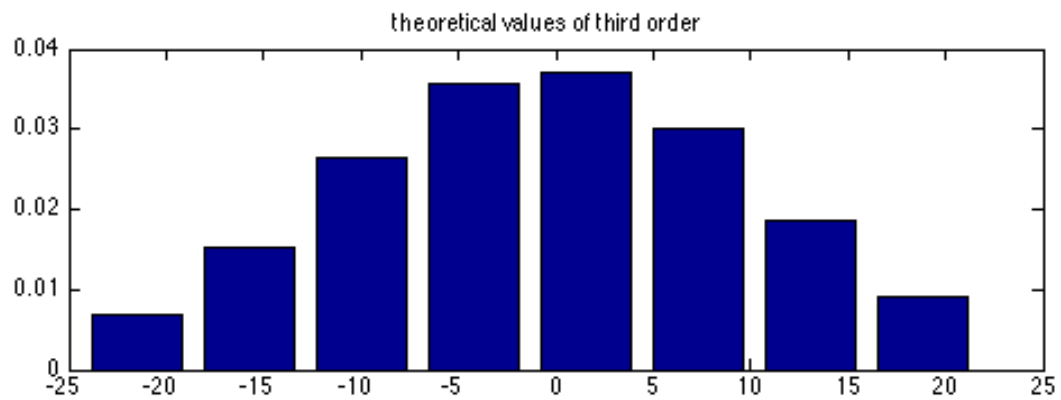
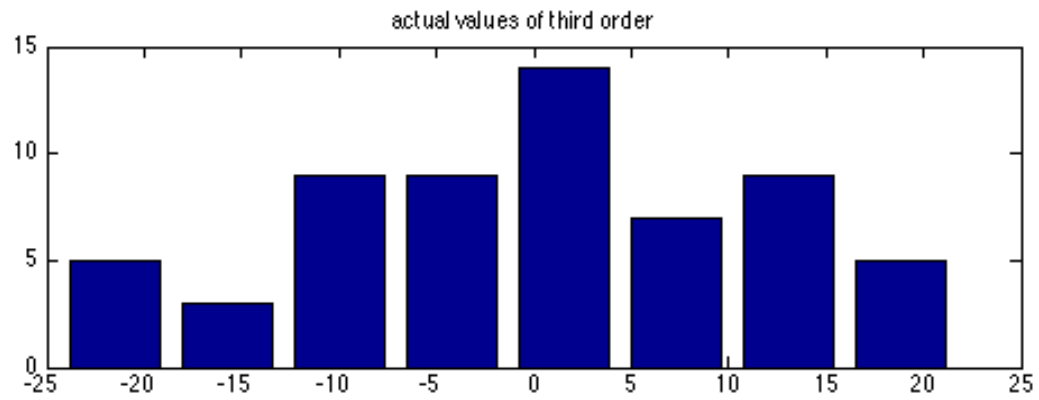
#3: Third-order polynomial fit

Checking if residuals are normally distributed:

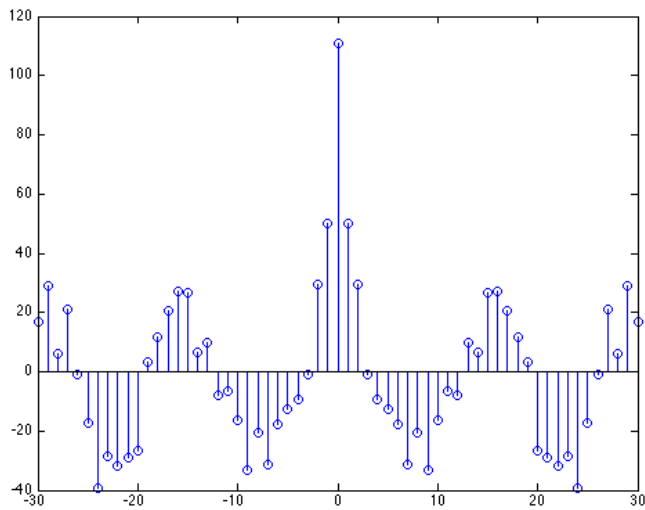
Calculated chi-squared value is 22.0441; critical chi-squared value is 27.5871

Calculated value < critical value means the null hypothesis cannot be rejected

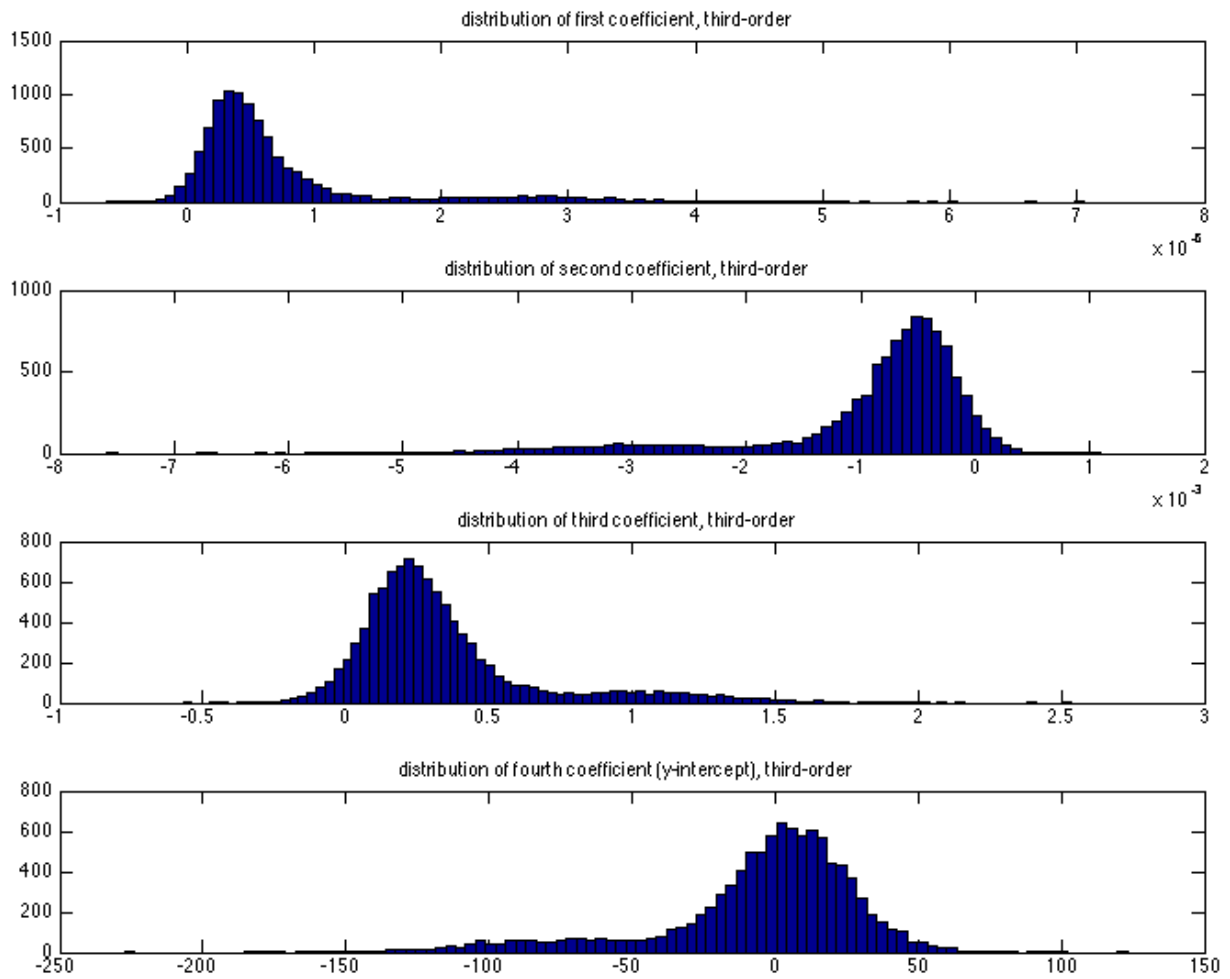
Thus, residuals are normally distributed



Autocorrelation analysis of residuals:



From this plot (lag = 20 days), the residuals for the third-order fit do seem to be auto-correlated because the residuals are symmetrical and not random.



The long tails for the bar graphs of the coefficients suggest that there are influential outlier points for the third-order polynomial.