

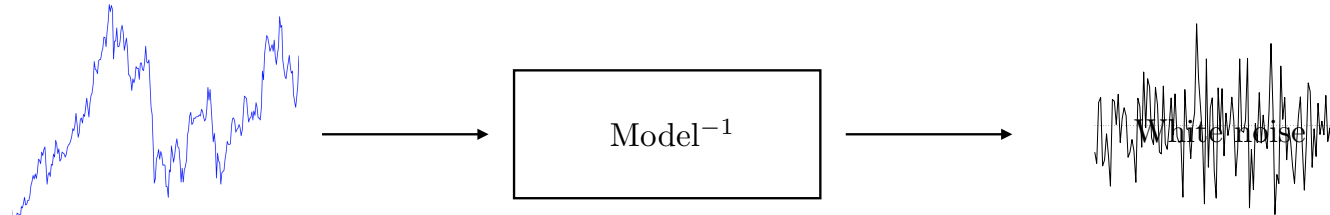


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# **Residual analysis**

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## Residual analysis

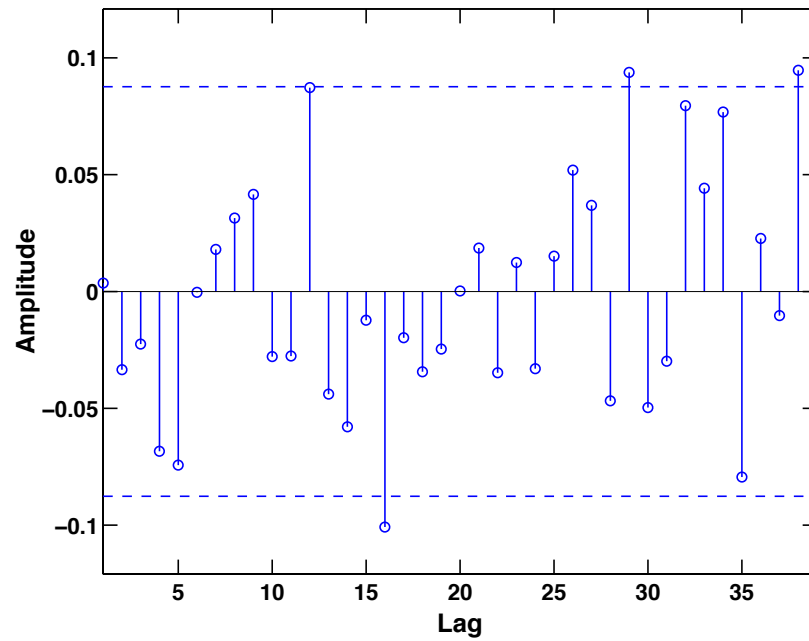


How can one determine if a sequence is "reasonably" white? In this course, we will examine a few different statistical test, namely

- Testing the ACF and the PACF
  - Plot and eyeball!
  - The Box-Pierce test
  - The Ljung-Box-Pierce test
  - The McLeod-Li test
  - The Monti test
- Testing the cumulative periodogram
- Testing sign changes

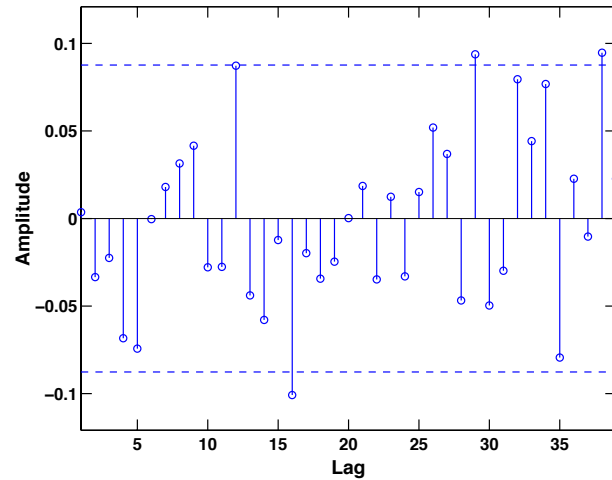
Important: *The first set of tests all assume that the ACF is Gaussian distributed, and may be unreliable if this assumption is violated!*

## Testing the ACF and the PACF



Remember that it is 95% confidence intervals; 1 in 20 may be slightly outside.

## Testing the ACF and the PACF



The Box-Pierce  $Q$ -statistic tests the hypothesis that the initial ACF estimates are not significantly different from zero, forming the  $Q$ -statistic as

$$Q = N \sum_{\ell=1}^K \hat{\rho}_{\hat{\varepsilon}}^2(\ell)$$

where  $K$  is the number of considered correlations. Thus, with significance  $\alpha$ , one may (asymptotically) reject the hypothesis that the residual is white if

$$Q > \chi_{1-\alpha}^2(K)$$

where  $\chi_{1-\alpha}^2(K)$  is the  $\alpha$ -quantile of the  $\chi^2$ -distribution with  $K$  degrees of freedom.

## Testing the ACF and the PACF

To better handle the small sample case, the Ljung-Box-Pierce  $Q$ -statistic tests introduce a scaling, instead forming

$$Q^* = N(N+2) \sum_{\ell=1}^K \frac{\hat{\rho}_{\hat{e}}^2(\ell)}{N-\ell}$$

One may also examine the squared residual and form the McLeod-Li test,

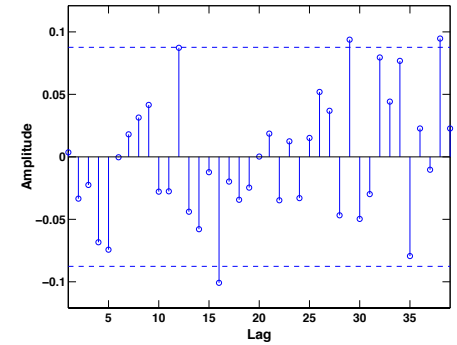
$$Q_2^* = N(N+2) \sum_{\ell=1}^K \frac{\hat{\rho}_{\hat{e}^2}^2(\ell)}{N-\ell}$$

One may instead test the PACF, forming the Monti test

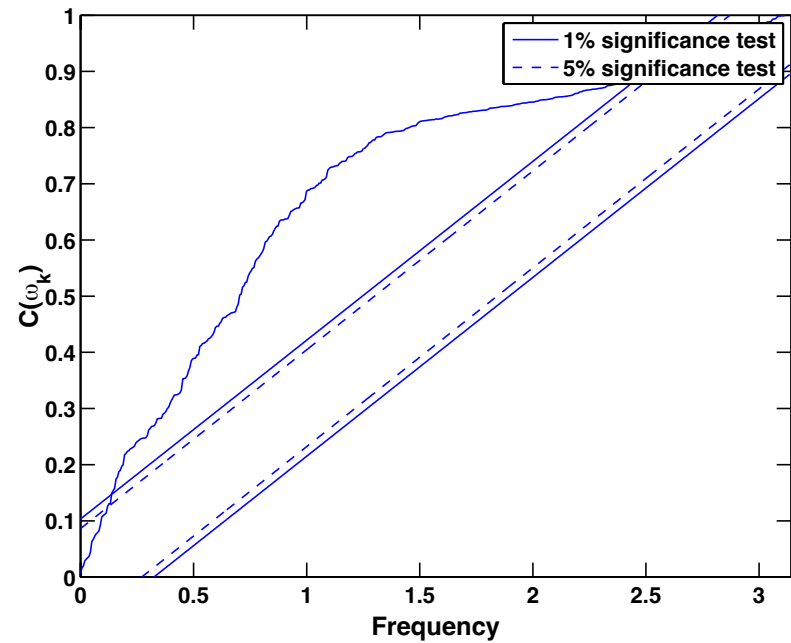
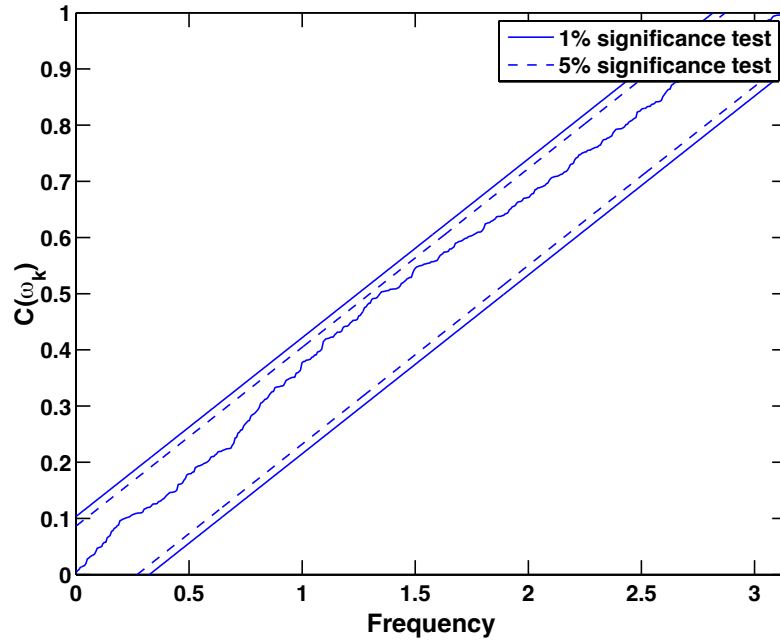
$$Q_M = N(N+2) \sum_{\ell=1}^K \frac{\hat{\phi}_{\ell,\ell}^2}{N-\ell}$$

The test statistic  $Q^*$ ,  $Q_2^*$ , and  $Q_M$  are all  $\chi^2(K)$  for a white signal. In general, the McLeod-Li test depends strongly on the Gaussian assumption and is often unreliable.

These tests are available using the functions `lbpTest`, `mlTest`, and `montiTest`.



## Testing the cumulative periodogram



As the (theoretical) spectrum of a white noise is flat, one can compare how well the sum of the periodogram, up to a given frequency  $\omega_k$ , follows a straight line:

$$\phi_c(\omega_k) = \frac{\sum_{\ell=1}^k \hat{\phi}_y(\omega_\ell)}{N \hat{\sigma}_y^2} = \frac{\frac{1}{N} \sum_{\ell=1}^k \left| \sum_{t=1}^N y_t e^{-i\omega_\ell t} \right|^2}{N \hat{\sigma}_y^2} = \frac{\sum_{\ell=1}^k \hat{\phi}_y(\omega_\ell)}{\sum_{\ell=1}^N \hat{\phi}_y(\omega_\ell)}$$

The corresponding confidence interval is  $\pm K_\alpha / \sqrt{[N - 1/2]}$ , above and below the theoretical line. The coefficient  $K_\alpha$  determines the  $(1 - \alpha)$  probability that  $y_t$  is white noise.

This test is available using the function `plotCumPer`.

$\alpha$	0.01	0.05	0.10	0.25
$K_\alpha$	1.63	1.36	1.22	1.02

## Testing sign changes

One can also examine the number of times the signal change sign,  $P$ . For a white signal, this should be about every second sample

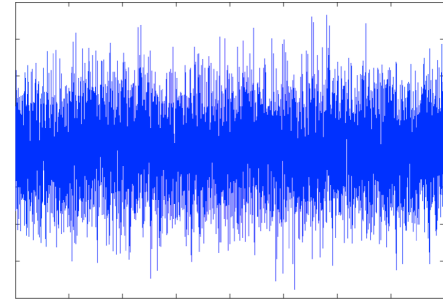
$$P \in \mathcal{B}\left(N-1, \frac{1}{2}\right)$$

For large data sets,  $P$  may be well approximated as being Normal distributed

$$P \in \mathcal{N}\left(\frac{N-1}{2}, \frac{N-1}{4}\right)$$

The test is available in the provided Matlab function `countSignChanges`.

The above discussed tests have all been collected in the function `whitenessTest`.



## Is this normal?

One may use the estimated skewness (asymmetry) and kurtosis (peakiness) to determine if an estimated distribution is Gaussian.

$S$  = skewness = 3rd moment /  $\text{std}^3$  (This is 0 for a Gaussian).

$K$  = kurtosis = 4th moment /  $\text{std}^4$  (This is 3 for a Gaussian).

The Jarque-Berra test is formed as

$$\gamma_{JB} = \frac{N}{6} \left( \hat{S}^2 + \frac{1}{4}(\hat{K} - 3)^2 \right)$$

which, for a Gaussian, is  $\chi^2$  distribution with two degrees of freedom. The test is provided in the function `jbtest`.

For small samples, the D'Agostino-Pearson's  $K^2$  test is often preferable. It is provided in the function `dagostinoK2test`.

To get a visual alternative, one may use the functions `normplot` or `plotNTdist`.

