

## Are those processes stationary and ergodic ?

Process 1 :

Independent draws at each point of time  $t$  from a standard normal distribution

$$\{\varepsilon_t\} \quad \varepsilon_t \sim N(0,1) \quad \text{for all } t=1, 2, \dots$$

↑  
used as a notation for  
stochastic process

Check  $E(\varepsilon_t), \text{Var}(\varepsilon_t), \text{Cov}(\varepsilon_t, \varepsilon_{t-j})$  for all  $j$

Process 2 :

$$\{Y_t\} \quad Y_t = Y_{t-1} + \varepsilon_t \quad \text{where } \varepsilon_t \sim N(0,1)$$

27

## How can we test for serial independence ?

serial independence  $\Rightarrow \text{Cov}(Z_t, Z_{t-j}) = E((Z_t - E(Z_t))(Z_{t-j} - E(Z_{t-j}))) = 0$  for all  $j \neq 0$

For a stationary process :  $\text{Cov}(Z_t, Z_{t-j}) = E((Z_t - \mu)(Z_{t-j} - \mu)) = \gamma_j = 0$  for all  $j \neq 0$

$$\text{Var}(Z_t) = E(Z_t - \mu)(Z_t - \mu) = \gamma_0$$

Equivalent:

$$\text{Correlation : } \rho_{XY} = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)}\sqrt{\text{Var}(Y)}} \quad -1 \leq \rho_{XY} \leq 1$$

$$\text{Autocorrelation : } \rho_{Z_t, Z_{t-j}} = \frac{\text{Cov}(Z_t, Z_{t-j})}{\sqrt{\text{Var}(Z_t)}\sqrt{\text{Var}(Z_{t-j})}}$$

$$\text{for stationary process } \rho_{Z_t, Z_{t-j}} = \frac{\gamma_j}{\sqrt{\gamma_0}\sqrt{\gamma_0}} = \frac{\gamma_j}{\gamma_0} = \rho_j$$

serial independence

No linear predictability

28

## How can we test for serial independence ?

If stochastic process is weakly stationary and ergodic

$$\hat{\mu} = \frac{1}{T} \sum_{t=1}^T z_t \quad \text{consistent estimator for}$$

$$\hat{\gamma}_0 = \frac{1}{T} \sum_{t=1}^T (z_t - \hat{\mu})^2 \quad \text{consistent estimator for}$$

$$\hat{\gamma}_j = \frac{1}{T-j} \sum_{t=j+1}^T (z_t - \hat{\mu})(z_{t-j} - \hat{\mu}) \quad \text{consistent estimator for}$$

$$\hat{\rho}_j = \frac{\hat{\gamma}_j}{\hat{\gamma}_0} \quad \text{consistent estimator for}$$

Compute  $\hat{\rho}_j$  from data

if serial independence (no linear predictability)

$\hat{\rho}_1, \hat{\rho}_2, \hat{\rho}_3, \dots$  should be close to zero

29

## How can we test for serial independence?

Assume: computation of  $\hat{\rho}_1$  yields 0.8 for  $T=1000$

$\Rightarrow$  would you accept hypothesis that  $\rho_1 = 0$  ?

Computation of  $\hat{\rho}_1$  yields -0.003 for  $T=1,000$

would you accept hypothesis that  $\rho_1 = 0$  ?

How likely or unlikely is the outcome assuming the (Null) Hypothesis is true?

You can commit 2 type of errors

type 1 : reject null hypothesis despite that it is true

type 2 : do not reject null hypothesis despite that it is false

Significance level ( $\alpha$ ) Maximal probability of committing type 1 error you are willing to accept in your decision to reject or maintain your null hypothesis

30

## Remember the paradigm of statistical testing (Neyman-Person)

$\alpha$  (significance level) = 0.001 (0.1%)

Assume hypothesis is true (null hypothesis). If the probability that outcome occurs ('under the null hypothesis') is less than 0.1% you reject the null hypothesis.

$\alpha$  (significance level) = 0.2 (20 %)

Assume hypothesis is true (null hypothesis). If the probability that outcome occurs ('under the null hypothesis') is less than 20 % you reject the null hypothesis

31

## Remember the paradigm of statistical testing (Neyman-Person)

- a) Fix significance level  $\alpha$
- b) Find a test statistic (computable from data) with known distribution assuming

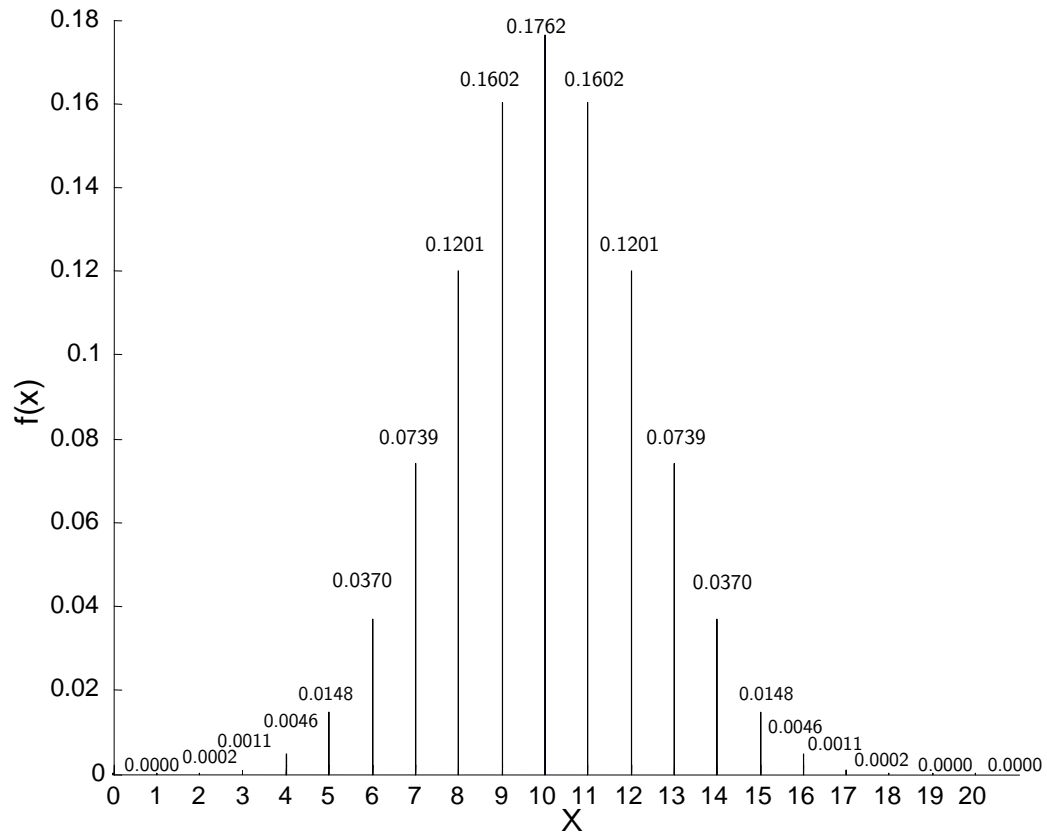
Null-Hypothesis is true

Under the Null-Hypothesis the distribution of the test statistic is...(Normal or Chi-square or Binomial or ....)

- c) From a) and b) determine acceptance and rejection regions

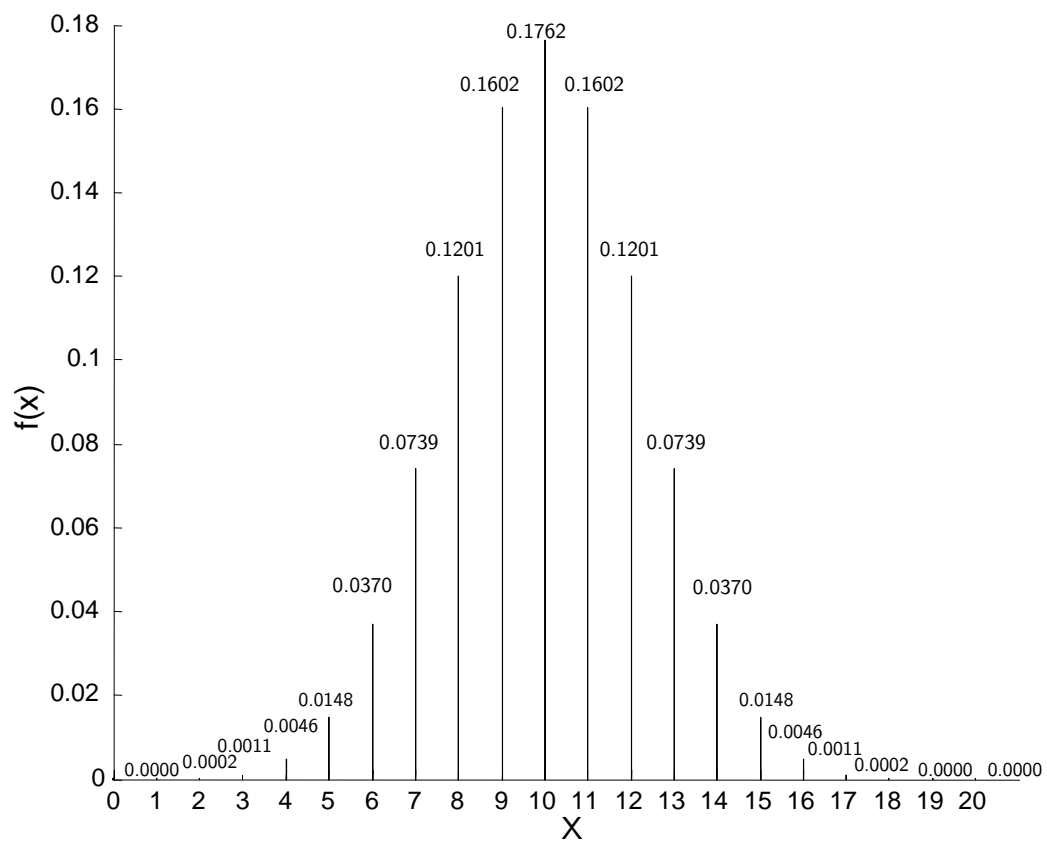
32

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33

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34