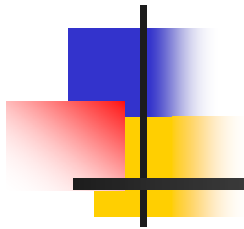


Introduction to Discrete Random Process



**87203 – Multimedial Signal
Processing 1st Module**

Politecnico di Milano –
Polo regionale di Como



Agenda

- Ex1: Bernoulli Random Sequence
- Ex2: Gaussian Random Sequence
- Ex3: Jointly distributed random sequences
- Ex4: Stationary vs ergodicity



Exercise 1 (1/6)

- Goal: Generate i.i.d. Bernoulli sequence of 1000 samples with arbitrary p .

$$f_x(x) = \begin{cases} 0 & \text{with probability } p \\ 1 & \text{with probability } 1-p \end{cases}$$

- A DTRS is a sequence of random variables $x(n)$

$$f_{x(n)}(x, n) = \sum_{k=-\infty}^{\infty} p_{x(n)}(x_k, n) \delta(x - x_k)$$



Exercise 1 (2/6)

- A DTRS is independent identically distributed when:

- the pdf does not depend on the time index

$$f_{x(n)}(x, n) = f_x(x)$$

- the random variables at different time instant n, m are independent

$$f_{x_j x_k}(x_j, x_k) = f_{x_j}(x_j) \cdot f_{x_k}(x_k)$$

- Hint: A i.i.d. sequence is generated by taking independent sample of the pdf.

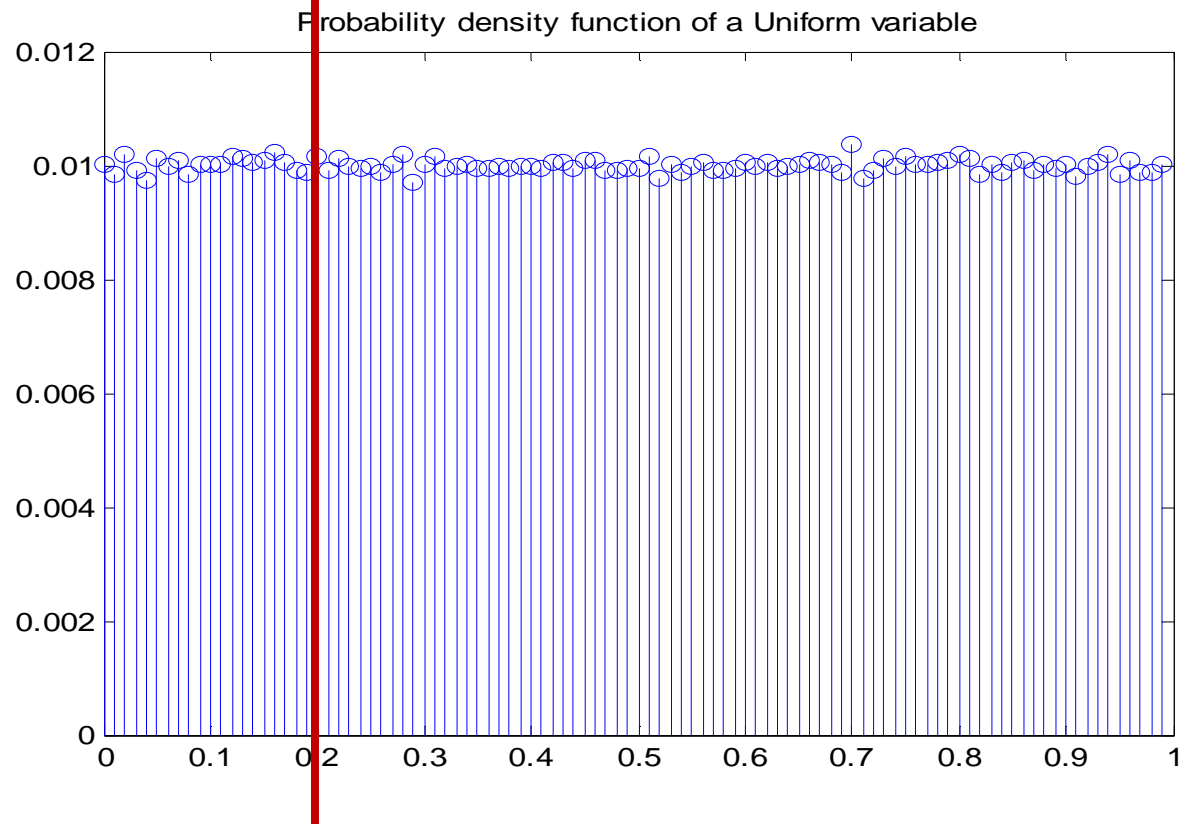


Exercise 1 (3/6)

- Pseudocode:
 - generate a iid sequence of 1000 samples with uniform distribution on the interval (0.0 , 1.0).
 - `y = rand(1000,1);`
 - assign value 0 at the index of each sample that is $< p$
 - `x2 = zeros(length(y),1);`
 - `p = 0.2;`
 - `x2(y < p) = 0 ;`
 - assign value 1 at the index of each sample that is $\geq p$
 - `x2(y \geq p) = 1 ;`

Exercise 1 (4/6)

- Let's see pdf of a discrete (100 possible set of values) uniform variable in the interval (0 , 1)





Exercise 1 (5/6)

- Compute the expected value and variance of the random variable x of which we have N realizations:

- mean: $E[x] = \eta_x = \int_{-\infty}^{\infty} x f_x(x) \delta(x)$

- `m=mean(x);`

- `ms=(1/N)*sum(x);`

$$\eta_x = 1 * p + 0 * (1 - p)$$

- variance: $\sigma_x^2 = E[(x - \eta_x)^2]$

- `sigma=var(x);`

- `vars=(1/N-1).*sum((x-ms)^2);`

$$\sigma_x^2 = (1 - p) * p$$



Exercise 1 (6/6)

- Matlab provides the functions:
 - “`x=rand(M,N)`” that produces a M-by-N matrix with random entries, chosen from a uniform distribution on the interval (0.0,1.0).
 - “`m=mean(x)`” that compute the mean value:
 - For vectors, “m” is the mean value of the elements in “x”
 - For matrices, “m” is a row vector containing the mean value of each column.
 - “`s=var(x)`” that compute the variance:
 - For vectors “s” returns the variance of the values in “x”.
 - For matrices, “s” is a row vector containing the variance of each column of “x”.



Exercise 2 (1/5)

- Goal: Generate i.i.d. Gaussian sequence of 1000 samples with arbitrary mean and variance.

$$f_x(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} e^{-\frac{(x-\eta_x)^2}{2\sigma_x^2}}$$

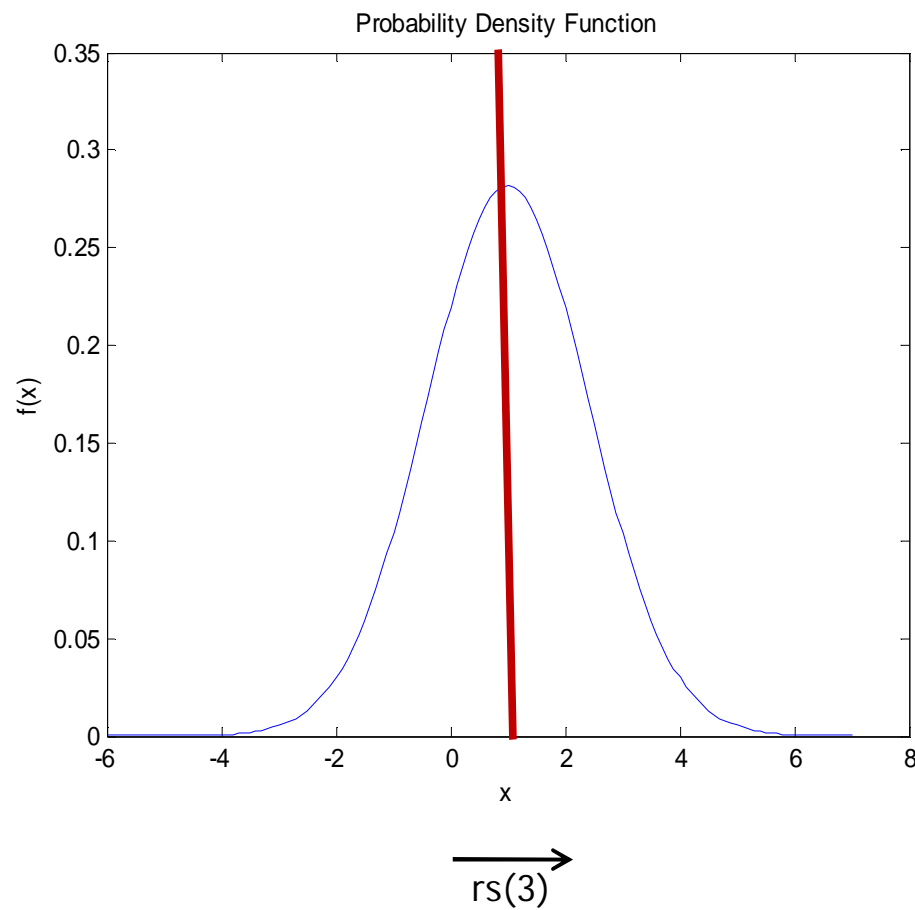
- Hint: A i.i.d. sequence is generated by taking independent sample of the pdf.



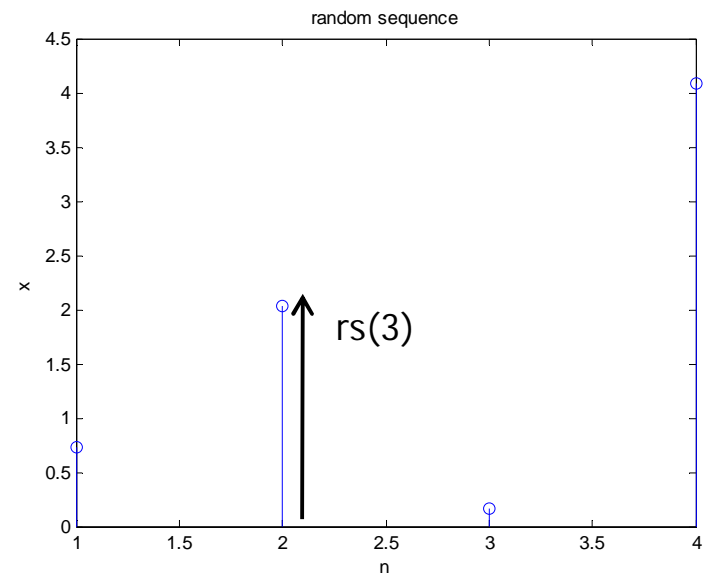
Exercise 2 (2/5)

- Pseudocode:
 - generate a iid sequence of 1000 samples with normal distribution
 - $x = \text{randn}(1000,1);$
 - set variance = s
 - $x = \text{sqrt}(s) * \text{randn}(1000,1);$
 - set mean = m
 - $x = \text{sqrt}(s) * \text{randn}(1000,1) + m;$
- N.B. For a gaussian distribution, 99,7% of possible value are inside the interval $m \pm 3s$

Exercise 2 (3/5)



$rs = [0.7360 \quad 2.0264 \quad 0.1680 \quad 4.0875]$



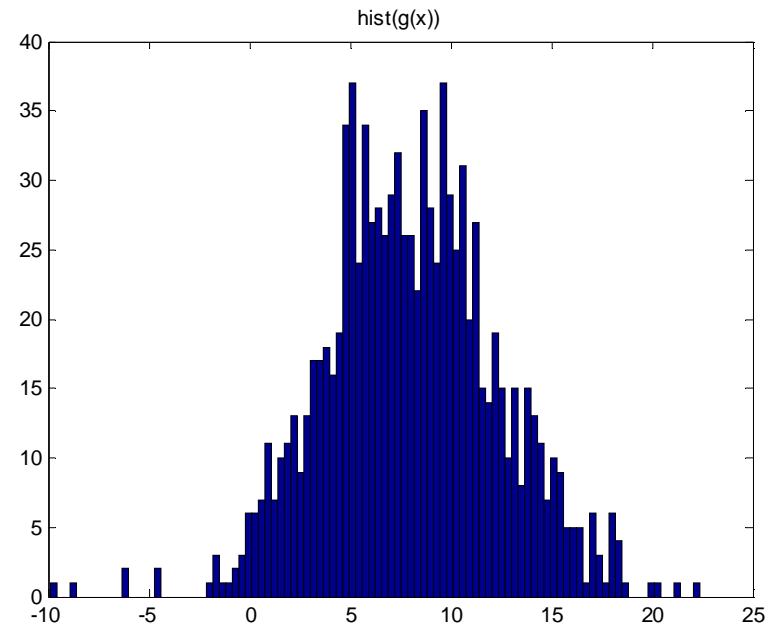
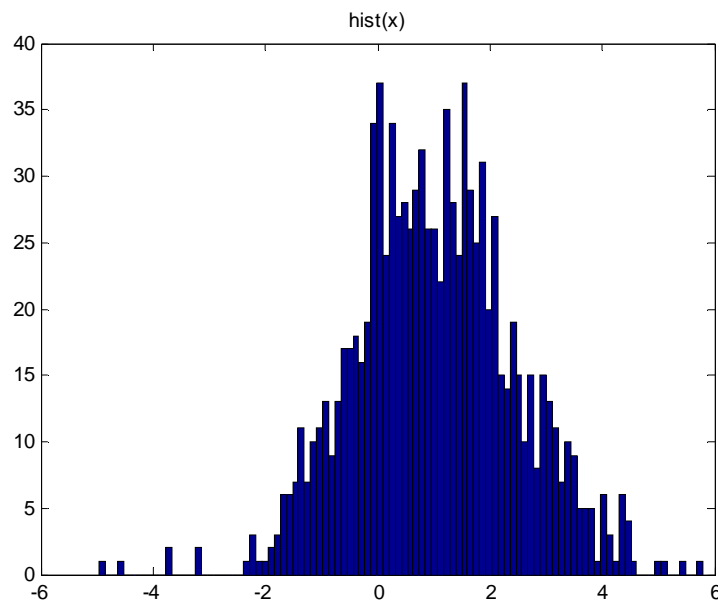
Exercise 2 (4/5)

■ Hint: Remember:

■ if x is a gaussian variable $x \approx N(\eta_x, \sigma_x^2)$

then $g(x) = ax + b$ is a gaussian variable

$$y = g(x) \approx N(a\eta_x + b, a^2\sigma_x^2)$$





Exercise 2 (5/5)

- Matlab provides the functions:
 - “`x=randn(M,N)`” that produces a M-by-N matrix with random entries, chosen from a normal distribution with mean zero, variance one and standard deviation one.
 - “`n=hist(x,M)`” that bins the elements of “x” into M equally spaced containers and returns the number of elements in each container “n”. If “x” is a matrix, “hist” works down the columns.



Exercise 3 (1/2)

- Goal: Generate two jointly distributed Gaussian random sequences.
 - Two DTRS $x(n)$ and $y(n)$ are described by the a joint probability function

$$F_{x(n)y(m)}(x, n, y, m) = P\{x(n) \leq x \text{ and } y(m) \leq y\}$$

- Assuming that the joint probability function is independent on the time index, it is a function of two variables x and y



Exercise 3 (2/2)

- Pseudocode:
 - generate a iid sequence of 1000 samples $x(n)$
 - generate a iid sequence of 1000 sample $z(n)$
 - consider the sequence $y(n)$ defined as:
 - $y(n) = x(n) + z(n);$

- N.B. $\{x(n)\}$ and $\{y(n)\}$ are statistically dependent: the knowledge of $f_x(x)$ and $f_y(y)$ is not enough to describe the joint pdf $f_{x,y}(x,y)$.



Exercise 4 (1/8)

- A DTRS is stationary if its statistical characterization is not affected by a shift in the data sequence:

$$F_{x(n)}(x, n) = F_{x(n+m)}(x, n+m) = F_x(x) \quad \forall n, m$$

- A DTRS is Wide-Sense Stationary if:

- mean value $E[x(n)] = \eta(n) = \eta \quad \forall n$
- autocorrelation

$$r_x(m) = E[x(n)x^*(n-m)] = E[x(n+m)x^*(n)]$$



Exercise 4 (2/8)

- A WSS DTRS is also ergodic when time averages are equivalent to ensemble averages:
 - mean value $\langle x(n) \rangle = E[x(n)]$
 - autocorrelation $\langle x(n) x^*(n-m) \rangle = r_x(m)$
- N.B.: A WSS sequence is not necessary ergodic



Exercise 4 (3/8)

- Goal: Generate R realization of a WSS random sequence (N samples) that is also ergodic.
 - Compute ensemble averages
 - Compute time averages
 - Demonstrate that the sequence is ergodic
- Hint: Matlab provides the function `"[r lag]=xcorr(x)"` that produces an estimate of the autocorrelation of the stationary sequence "x". "lag" is the vector of lag indices.



Exercise 4 (4/8)

- Pseudocode:
 - Consider an iid sequence of $\{z(n)\}$ with gaussian zero-mean distribution
 - $z = \text{sqrt}(\text{varz}) * \text{randn}(1000, 1);$
 - Consider the sequence $\{x(n)\}$ generated by:

$$x(n) = \rho x(n-1) + z(n)$$

- $x = \text{filter}(1, [1 \ -\rho], z);$
- Compute R sequence $\{x(n)\}$ of N samples each.
 - $z = \text{sqrt}(\text{varz}) * \text{randn}(N, R);$



Exercise 4 (5/8)

- Pseudocode (continued):
 - Compute ensemble averages:
 - $mR = \text{mean}(x(:,50));$
 - $vR = \text{var}(x(:,50))$
 - $[rR \text{ lag}R] = \text{xcorr}(x(:,50));$
 - Compute time averages
 - $mN = \text{mean}(x(1,:))$
 - $vN = \text{var}(x(1,:))$
 - $[rN \text{ lag}N] = \text{xcorr}(x(1,:));$
 - Demonstrate that:
 - $mN = mR;$ $vN = vR;$ $rN = rR;$



Exercise 4 (6/8)

- Goal: Generate R realization of a WSS random sequence (N samples) that is not ergodic.
 - Compute ensemble averages
 - Compute time averages
 - Demonstrate that the sequence is NOT ergodic



Exercise 4 (7/8)

- Pseudocode:
 - Let a be a random number normally distributed:
 - $a = \text{randn}(R,1);$
 - One realization of the process is a constant horizontal time path:
 - $x = \text{ones}(R,N);$
 - for $i=1:R$
 - $x(i,:) = a(i) .* x(i,:);$
 - end



Exercise 4 (8/8)

- Pseudocode (continued):
 - Compute ensemble averages:
 - $mR = \text{mean}(x(:,50));$
 - $vR = \text{var}(x(:,50))$
 - $[rR \text{ lag}R] = \text{xcorr}(x(:,50));$
 - Compute time averages
 - $mN = \text{mean}(x(1,:))$
 - $vN = \text{var}(x(1,:))$
 - $[rN \text{ lag}N] = \text{xcorr}(x(1,:));$
 - Demonstrate that:
 - $mN \neq mR;$ $vN \neq vR;$ $rN \neq rR;$