### **Some Well-known Distributions**

이 상화



### Geometric Distribution

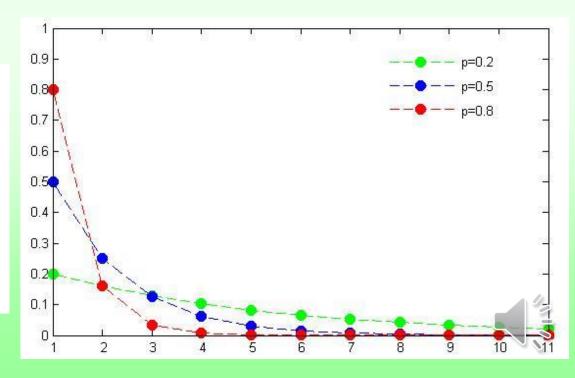
- $\square X$ : number of trials until the 1<sup>st</sup> success
  - p: probability of success for a trial
- ☐ Applications
  - Network server access trials until success

$$P(X = n) = p(1-p)^{n-1}$$

$$P(X > n) = (1-p)^n$$

$$\text{Mean } \mu = \frac{1}{p}$$

$$\text{Variance } \sigma^2 = \frac{1-p}{p^2}$$

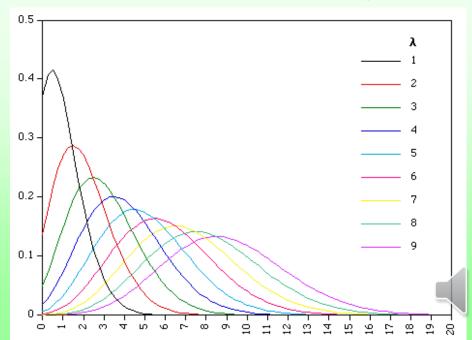


### Poisson Distribution

- $\square X$ : number of occurrences in a time interval
  - $\lambda$ : mean, variance
- ☐ Applications
  - Network server access trials for a time interval
  - Service system queues: Bank counter, Tollgate

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

$$P_X(k) = \frac{e^{-(\lambda t)} * (\lambda t)^k}{k!}$$



# **Exponential Distribution**

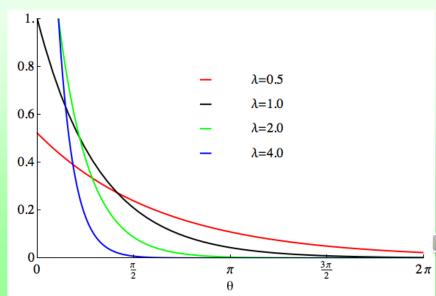
- $\square X$ : interval (or distance) between occurrences
  - $\lambda$ : rate (mean of counts in a unit interval)
- ☐ Applications
  - Time interval between successive occurrences of independent identical distributed events.
    - > EX: time interval between guests in the bank

$$f(x) = \lambda e^{-\lambda x}$$

$$f(x) = \lambda e^{-\lambda x}$$
$$F(x) = 1 - e^{-x\lambda}$$

$$E(X) = 1/\lambda$$

$$V(X) = 1/\lambda^2$$





# Erlang-r Distribution

#### $\square X$ : interval between r successive occurrences

- $\lambda$ : rate (mean of counts in a unit interval)
- r-1 convolutions of exponential distributions

#### ■ Applications

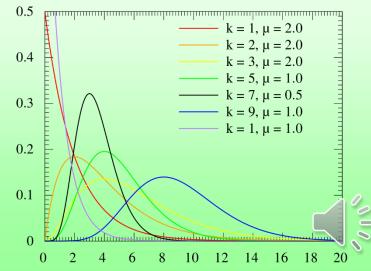
■ Time interval between *r* successive occurrences of independent identical distributed events.

> EX: time interval between guests in the bank

$$P(X > x) = \sum_{k=0}^{r-1} \frac{e^{\lambda x} (\lambda x)^k}{k!} = 1 - F(x)$$

Now differentiating F(x):

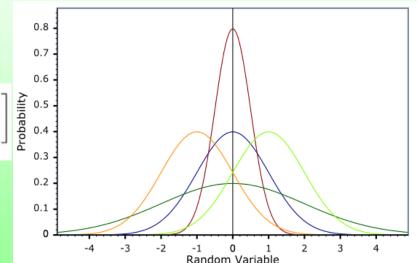
$$f(x) = \frac{\lambda^r x^{r-1} e^{-\lambda x}}{(r-1)!}$$
 for  $x > 0$  and  $r = 1, 2, ...$ 



## Gaussian Distribution (1/2)

- $\square X$ : sum of independent identical distributed random variables
  - $X = X_1 + X_2 + X_3 + \dots X_n \quad (n \rightarrow \infty)$
  - Convolutions of pdf of X<sub>i</sub>
- ☐ Applications
  - General distribution for any signals
  - Noise

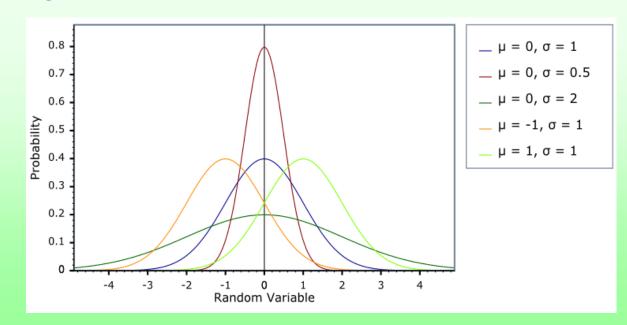
$$\mathcal{N}(x \; ; \; \mu, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left[-\frac{1}{2}(x-\mu)^2/\sigma^2\right]_{\substack{\frac{\lambda}{\alpha} \\ \alpha \\ 0.4}}^{\frac{\lambda}{\alpha}} {}_{0.4}$$





## Gaussian Distribution (2/2)

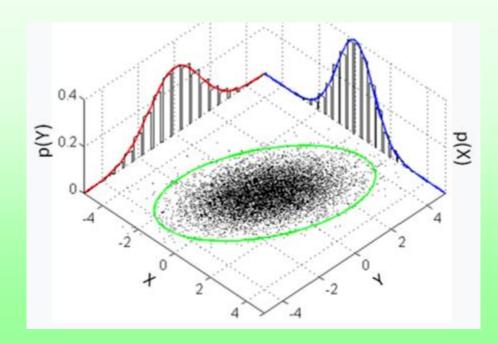
- ☐ Single mode
  - There is one value with the maximum probability.
- ☐ Symmetric at the mean
  - Maximum probability => mean
- ☐ Finding the mean is the best for Gaussian RV.

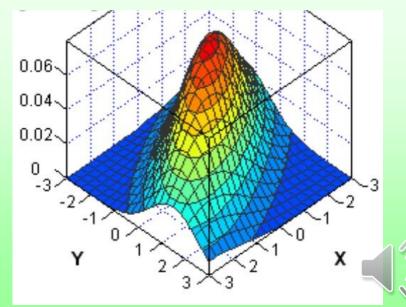




## Multi-variate Distributions (1/3)

- ☐ Two or more random variables are considered simultaneously.
  - Influenced or dependent on each other
- $\square$  Discrete RVs: P(X=x, Y=y)
- $\square$  Continuous RVs: Joint pdf: f(x, y)



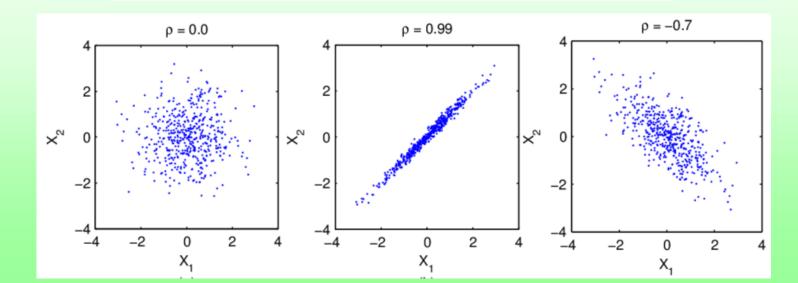


## Multi-variate Distributions (2/3)

#### ☐ Correlation of 2 random variables

- Similarity measure of two random variables
- Normalized measure: -1 ~ +1

$$\rho_{X,Y} = \frac{cov(X,Y)}{\sigma_X \sigma_Y} = \frac{E[(X - \mu_X)(Y - \mu_Y)]}{\sigma_X \sigma_Y}$$



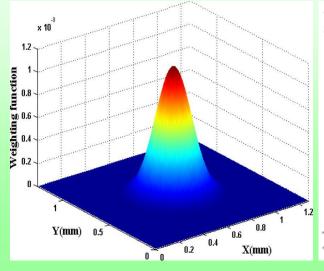


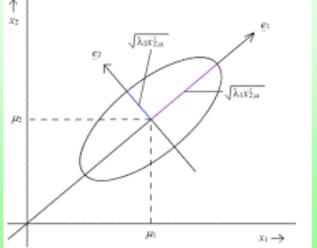
## Multi-variate Distributions (3/3)

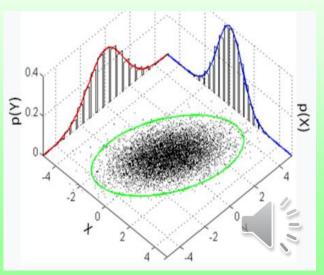
#### ☐ Bivariate Gaussian

Joint 2 Gaussian RVs

$$f_{XY}(x, y; \sigma_X, \sigma_Y, \mu_X, \mu_Y, \rho) = \frac{1}{2\pi\sigma_X \sigma_Y \sqrt{1 - \rho^2}} \exp\left\{\frac{-1}{2(1 - \rho^2)} \left[\frac{(x - \mu_X)^2}{\sigma_X^2}\right] - \frac{2\rho(x - \mu_X)(y - \mu_Y)}{\sigma_X \sigma_Y} + \frac{(y - \mu_Y)^2}{\sigma_Y^2}\right]\right\}$$
(5-







# HW#11 (1/2)

- ☐ Find the correlation coefficients between color components in the images
  - G-R in RGB space
  - G-B in RGB space
  - R-B in RGB space
  - Y-U in YUV space
  - Y-V in YUV space
  - U-V in YUV space
- ☐ Due date
  - 11/18 (Wed.) 22:00



# HW#11 (2/2)

- ☐ Collect 10 colorful images
- ☐ Use RGB components
- ☐ Use YUV components
  - Any YUV format is OK

$$\begin{bmatrix} Y' \\ U \\ V \end{bmatrix} = \begin{bmatrix} 0.299 & 0.587 & 0.114 \\ -0.14713 & -0.28886 & 0.436 \\ 0.615 & -0.51499 & -0.10001 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$
$$\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1.13983 \\ 1 & -0.39465 & -0.58060 \\ 1 & 2.03211 & 0 \end{bmatrix} \begin{bmatrix} Y' \\ U \\ V \end{bmatrix}$$

