Lab 2

Probability review

Pairwise independence does not imply mutual independence.

- $X = Flip coin, Pr(X = heads) = \frac{1}{2}$
- Y = Flip Coin, Pr(Y = heads) = ½
- Z = Pr(exactly X or Y are heads (not both)) = ½

X	Υ	Z = XOR(X,Y)
0	0	0
1	0	1
0	1	1
1	1	0

X	Υ	Pr[X,Y]
0	0	1/4
1	0	1/4
0	1	1/4
1	1	1/4

X	Z	Pr[X,Y]
0	0	1/4
1	1	1/4
0	1	1/4
1	0	1/4

Pairwise independence does not imply mutual independence.

- Pr(X,Z) pairwise independence:
 - Pr(X=0,Z=0) = Pr(X=0)* Pr(Z=0) = 1/2 * 1/2 = 1/4
- NOT mutual independence
 - $Pr(X=1,Y=1,Z=0) = \frac{1}{4}$ where as $Pr(X=1)*Pr(Y=1)*Pr(Z=0) = \frac{1}{8}$

X	Υ	Z = XOR(X,Y)	Pr[X]*Pr[Y]*Pr[Z]	Pr[X,Y,Z]
0	0	0	1/8	1/4
1	0	1	1/8	1/4
0	1	1	1/8	1/4
1	1	0	1/8	1/4
0	0	1	1/8	0
1	0	0	1/8	0
0	1	0	1/8	0
1	1	1	1/8	0

Definitions

- Expectation
 - When X is a discrete random variable

$$lacksquare \sum i \cdot \Pr[X=i]$$

 When X is a continuous random variable, and f(x) is its probability density function

$$\bullet \int_{-\infty}^{\infty} x \cdot f(x) dx$$

Variance

$$egin{array}{ll} \circ \ Var[X] = Eigl[(X-E[X])^2igr] = Eigl[X^2igr] - E[X]^2igr] \end{array}$$

- o It measures how far your data spread out from their average value.
- Standard deviation

$$\circ \ \ \sigma_X = \sqrt{Var[X]}$$

Covariance and Correlation

Definitions:
$$Cov(X,Y)=E[(X-E[x])(Y-E[Y])]$$

$$=\sum_x\sum_y(x-E[X])(y-E[Y])p_{X,Y}(x,y)$$

$$Corr(X,Y)=\frac{Cov(X,Y)}{\sigma_X\sigma_Y}$$

From a linear algebra perspective: If X, Y are two random variables of zero mean, then the covariance $Cov[XY] = E[X \cdot Y]$ is the dot product of X and Y. The standard deviation of X is the length of X. The correlation is the cosine of the angle between the two vectors.

Covariance

Some properties:

$$Cov(X + c, Y) = Cov(X, Y)$$

$$Cov(aX + bY, Z) = a \cdot Cov(X, Z) + b \cdot Cov(Y, Z)$$

Exercise:

X and Y are two independent $\mathcal{N}(0,1)$ random variables and:

$$Z = 1 + X + XY^2$$

$$W = 1 + X$$

Find Cov(W,Z)

Discrete probability distributions

	PMF	Mean	Variance
Bernoulli (p)	1-p if k=0 p if k=1	p	p(1-p)
Binomial (n,p)	$\binom{n}{k}p^k(1-p)^{n-k}$	np	np(1-p)
Poisson (λ)	$rac{\lambda^k e^{-\lambda}}{k!}$	λ	λ
Geometric (p)	$(1-p)^{k-1}p$	$\frac{1}{p}$	$rac{1-p}{p^2}$

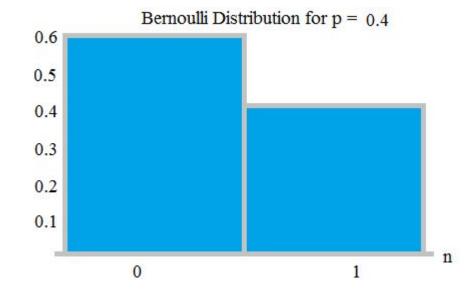
Bernoulli(p)

- Discrete
- Two outcomes: 0,1 | | fail, success

•
$$Pr(X = 1) = p = 1-q$$

•
$$Pr(X = 0) = q = 1-p$$

- Expectation:
 - E[x] = 1*p + (0)*(1-p) = p



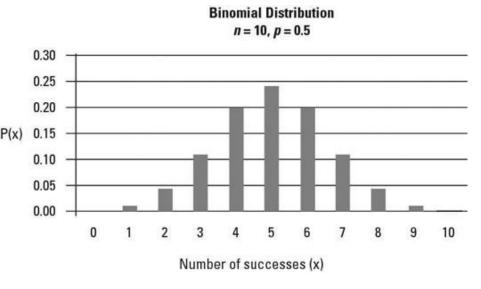
Binomial(n,p)

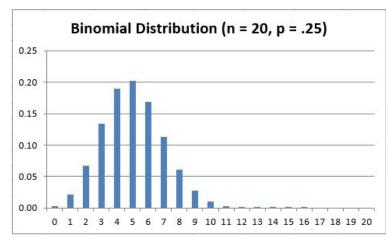
Example: Flip a coin 5 times, what is the probability of having exactly 2 heads?

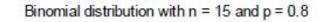
- Sample repetitively from a Bernoulli distribution n times
- Bernoulli = Binomial with n=1
- $ullet ext{Pr}[X=k] = inom{n}{k} p^k (1-p)^{n-k} \ ext{Expectation}$
 - - Use linearity of expectations
 - \circ Let X_i be the ith sample outcome from the Bernoulli distribution
 - $^{\circ} E[X_1 + ... X_n] = \sum_i E[X_i] = n \cdot p$

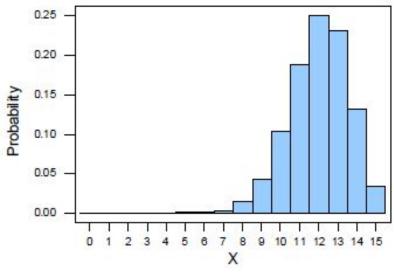
Binomial Distributions

$$\mathsf{PDF} \quad \Pr[X = k] = inom{n}{k} p^k (1-p)^{n-k}$$



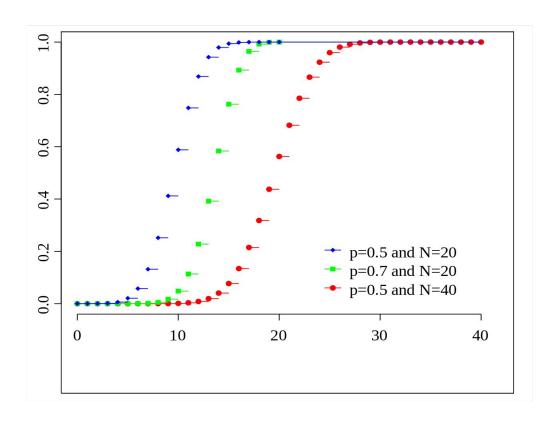






Binomial Distribution CDF

• CDF: $F_{\chi}(t) = \Pr(X \leq T)$



$Poisson(\lambda)$

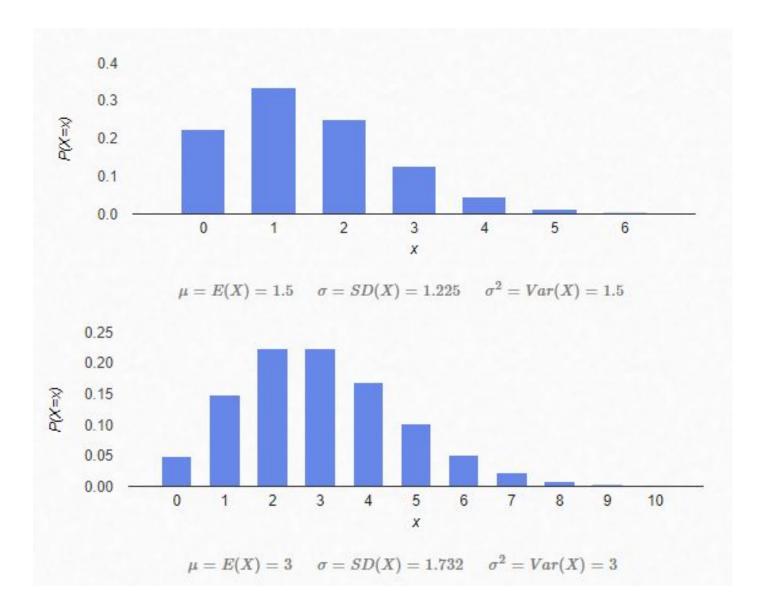
Poisson distribution is a probability distribution that is used to show how many times an event is likely to occur over a specified period

- Events are independent of each other.
- The occurrence of one event does not affect the probability another event will occur.
- The average rate (events per time period) is constant.
- Two events cannot occur at the same time.

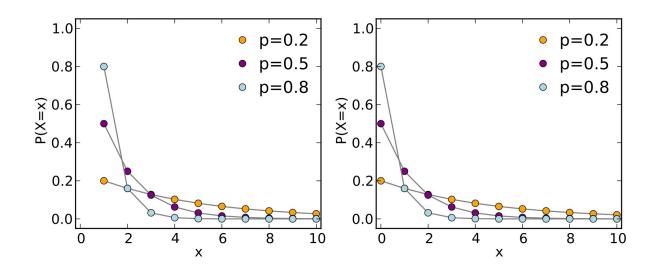
$$\Pr[X=k] = rac{\lambda^k e^{-\lambda}}{k!}$$

Poisson Example

- Example: You are "on call" at work. And every night between 12 and 8
 AM you receive about 6 calls from customers
 - Question: What is the probability you receive a call from 2-4 AM?
 - Definitions: X~P(lambda) where X is a random variable with a Poisson distribution, and "lambda" is the mean for the interval
 - 6 calls over an 8 hour window is about $\frac{3}{4}$ of a call every hour, our window is 2 hours, so our "lambda" value is 2*3/4 = 1.5
 - https://homepage.divms.uiowa.edu/~mbognar/applets/pois.html



Geometric(p)



- Discrete, closely related to Bernoulli
- Probability of N failures before a success
- $Pr(X = k) = (1 p)^{k-1}p^1$

Expected Value:

$$\sum_{k=1}^{\infty} k * (1-p)^{k-1}p$$

 Example: What is the probability I flip a coin 4 times before getting heads?

Geometric Expectation Derivation

1.
$$E[X] = \sum_{k=1}^{\infty} k * (1-p)^{k-1}p$$

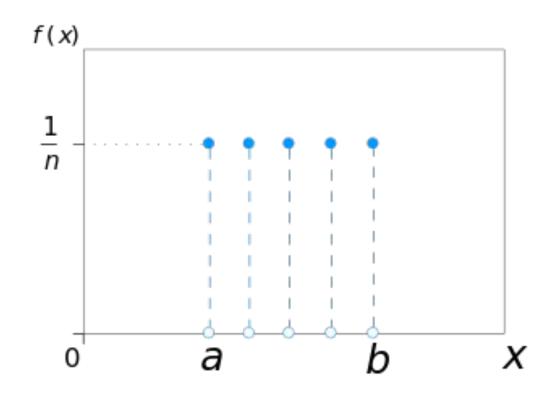
2. =
$$p * \sum_{k=1}^{\infty} k * (1-p)^{k-1} // \text{ take out p from summation}$$

3. =
$$p * \sum_{k=1}^{\infty} \frac{d}{dp} * -(1-p)^k$$
 // substitute derivative w.r.t p

4. =
$$p * -\frac{d}{dp} \sum_{k=1}^{\infty} (1-p)^k$$
 // well known series summation

5.
$$= -p * \frac{d}{dp} * \frac{1}{p} = -p \left(-\frac{1}{p^2} \right) = \frac{1}{p}$$

Uniform - Discrete

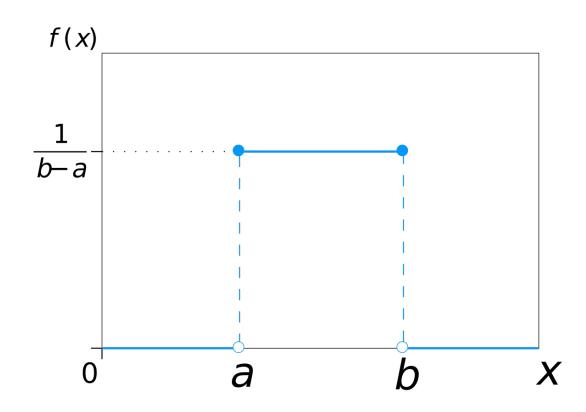


$$Var[x] = \frac{N^2 - 1}{12}$$

Continuous probability distributions

	PDF	Mean	Variance
Uniform (a,b)	$rac{1}{b-a}$ for $a \leq x \leq b$ 0 otherwise	(a+b)/2	$(b-a)^2/12$
Exponential (λ)	$\lambda e^{-\lambda x}$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$
Laplace (μ, b)	$\frac{1}{2b} \exp\left(-\frac{ x-\mu }{b}\right)$	μ	$2b^2$
Gaussian (μ, σ^2)	$\frac{1}{\sigma\sqrt{2\pi}}\exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$	μ	σ^2

Uniform - Continuous

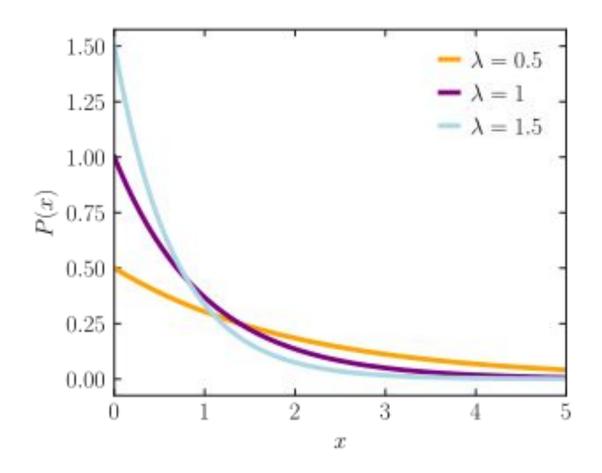


E[X] where P(X=x) = 1/N:

$$\frac{1}{b-a}$$

$$Var[x] = \frac{(b-a)^2}{12}$$

Exponential

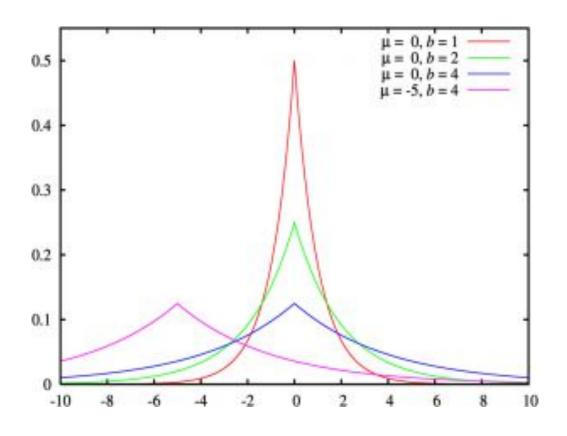


The exponential distribution is the probability distribution of the time between events in a Poisson point process.

- Parameterized by (λ)
- ullet PDF $\Pr[X=x]=\lambda e^{-\lambda x}$
- ullet Expected value $1/\lambda$
- Variance $1/\lambda^2$

Remark: Exponential distribution is not the same as the class of exponential families of distributions

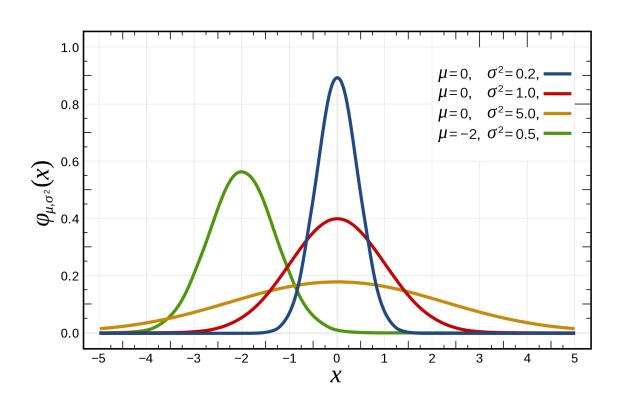
Laplace (μ, b)



- It can be thought of as two exponential distributions (with an additional location parameter) spliced together along the abscissa
- Parametrized by mu and beta:
 - Mu = Location (shift along the x axis)
 - B = scaling, maximum peak value

$$\Pr[X=x] = rac{1}{2b} \mathrm{exp}\left(-rac{|x-\mu|}{b}
ight)$$

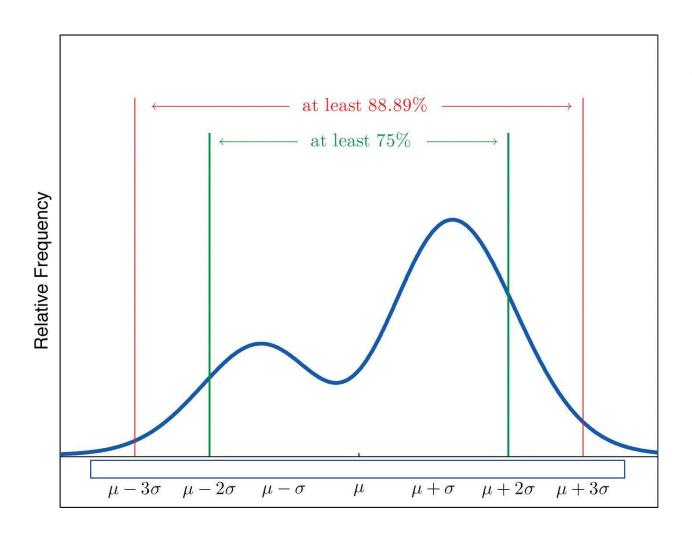
Gaussian – One dimension



$$\Pr[X=x] = rac{1}{\sigma\sqrt{2\pi}} \mathrm{exp}\left(-rac{\left(x-\mu
ight)^2}{2\sigma^2}
ight)$$

- Parameterized by
 - $\circ \mu$ = mean
 - $\circ \sigma$ = standard deviation
- It is important partly due to CLT.

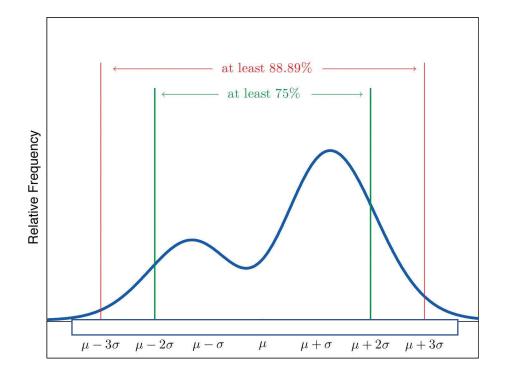
Chebyshev Inequality



$$P(|X - \mu| \ge k * \sigma) \le \frac{1}{k^2}$$

Poisson Chebyshev problem

- Let us say that X is modeled by a Poisson variable with mu/lambda/mean = 100
- What is the lower bound of P(75 < X < 125)?



Poisson Chebyshev solution

- Find μ and σ :
 - μ = 100
 - $\sigma = \sqrt{var} = \sqrt{100} = 10$
- $P(|X \mu| \ge k * \sigma) \le \frac{1}{k^2}$ is the P that we exist outside a certain SD
- What we want is the P we exist WITHIN a certain SD
- $P(75 < X < 125) = 1 P(|X 100| \ge 2.5 * 10) \ge 1 \frac{1}{2.5^2} = 0.84$

End

Practice Problem

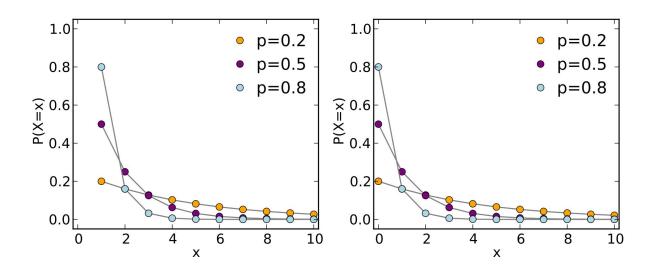
- Broken device, calling customer support
- 1% chance a human representative being available and picking up my call, or else robot will disconnect me

Question:

 What is the expected number of times I will have to call in before I talk to a human being?

Practice Solution

- Identify an analytical distribution:
 - Geometric in our case (fails until a success)



- Discrete, closely related to Bernoulli
- Probability of N failures before a success
- $P(X = k) = (1 p)^{k-1}p^1$
- Example: What is the probability I flip a coin 4 times before getting heads?

Practice Solution

- X = number of call attempts before a human picks up
- p = .01, q = 1-p
- $E[X] = \sum_{i=0}^{N} X_i * P(X_i)$
- = $1(.01) + 2(.99)^{1}(.01) + 3(.99)^{2}(.01) + ...$
- = $p/(1-q)^2$ = 1/p = 1/.01 = **100**