

# Poisson Distribution

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# Overview

- The Poisson (pwa - son) Distribution is used to determine the probability of some number of events occurring within a given span of time
- It can be derived from the Binomial Distribution in which a biased coin is flipped repeatedly
- In computer simulations, the random generation of Poisson distributed events can be simplified
- In a Poisson Process, events occur continuously and independently of each other with a variance equal to the mean

# Permutation

- Ordered arrangement of  $n$  distinct objects
- $n$  choices for first selection,  
 $n - 1$  choices for second selection,  
 $n - 2$  choices for third selection,  
[...],  
1 choice for last selection
- $n * (n - 1) * (n - 2) * [...] * 3 * 2 * 1$
- $n$  factorial
- $n!$

# Permutation subset

- Choose ordered subset  $k$  of  $n$  distinct objects
- Example: choose just 3 of 7 distinct objects
- 7 first choices, 6 second, 5 third
- $n * (n - 1) * \dots * (n - k + 1)$
- Example:  $7 * (7 - 1) * (7 - 3 + 1) = 7 * 6 * 5$
- Same as dividing  $n$  factorial by  $(n - k)$  factorial:  
$$\frac{n * (n - 1) * \dots * (n - k + 1) * (n - k) * \dots * 1}{(n - k) * \dots * 1}$$
- $n! / (n - k)!$
- Example:  $7! / (7 - 3)! = 7! / 4! = 7 * 6 * 5$

# Combinations vs. Permutations

- Permutations ordered  
abc, acb, bac, bca, cab, cba
- Combinations unordered: abc
- Choose unordered subset k of n distinct objects
- Example: choose 2 from set of 3 (abc), ordered
- $n! / (n - k)! = 3! / (3 - 2)! = 3! / 1! = 3 * 2 = 6$
- ab, ac, ba, bc, ca, cb
- Example: choose 2 from 3, unordered
- ab (ab and ba), ac (ac and ca), bc (bc and cb)
- Number of combinations same as permutations divided by number of ways to permute subset

# Combinations

- To get number of combinations (unordered)
  - First calculate the number of ways we can create an ordered subset of length  $k$  from  $n$  distinct objects
  - Then divide it by the number of ways to order  $k$
- $[ n! / ( n - k )! ] / k!$
- $n! / [ ( n - k )! * k! ]$
- $n$  choose  $k$   $\binom{n}{k}$
- Example: 3 choose 2
- $3! / [ ( 3 - 2 )! * 2! ] = 3! / [ 1! * 2! ] = 3$
- combinations of 2 from abc: ab, ac, bc

# Random Combinations

- Urn with three white balls numbered 1, 2, and 3
- Randomly draw two and paint them black
  - 6 Permutations: 1, 2; 1, 3; 2, 1; 2, 3; 3, 1; 3, 2
  - 3 Combinations:  $w_1 b_2 b_3, b_1 w_2 b_3, b_1 b_2 w_3$
- Randomly draw one and paint it black
  - 3 Permutations: 1; 2; 3
  - 3 Combinations:  $w_1 w_2 b_3, w_1 b_2 w_3, b_1 w_2 w_3$
- Randomly draw three and paint them black
  - 6 Permutations but just one combination:  $b_1 b_2 b_3$

# Balls to Coins

- Urn with three coins numbered 1, 2, and 3
- Coins in the urn all start tails up
- Randomly drawn coins are placed heads up
- Randomly draw two and place heads up
  - 6 Permutations: 1, 2; 1, 3; 2, 1; 2, 3; 3, 1; 3, 2
  - 3 Combinations:  $t_1 h_2 h_3$ ,  $h_1 t_2 h_3$ ,  $h_1 h_2 t_3$
- Same as white balls being painted black
- Replace 3 drawn coins with 1 coin flipped thrice

# Coin Flips

- Flip a coin thrice
  - zero heads and three tails (ttt)
  - one head and two tails (tth, tht, htt)
  - two heads and one tail (thh, hth, hht)
  - three heads and zero tails (hhh)
- Calculate using combinations
  - n choose k:  $n! / [(n - k)! * k!]$
  - $3! / [(3-0)! * 0!] + 3! / [(3-1)! * 1!] + 3! / [(3-2)! * 2!] + 3! / [(3-3)! * 3!]$
  - $1 + 3 + 3 + 1 = 8$  different ways

# Ordering

- Permutations ordered, combinations unordered
- To determine the number of ways we can flip a coin  $n$  times and get  $k$  heads, we use the formula for unordered combinations
- Strange because it looks like we are ordering
  - two heads and one tail (thh, hth, hht)
- What was combined going from balls to coins?
  - second head, third head: thh
  - third head, second head: thh

# Fair and Biased Coins

- Probability of heads on fair coin:  $p = 0.5$  (50%)
- Biased coins do not have 50% chance of heads
- Probability of head on biased coin:  $p = 0.9$
- Probability of tail:  $q = 1 - p = 1 - 0.9 = 0.1$
- Whether fair or biased, the number of ways of getting 2 heads in 3 coin flips is still the same:  
n choose k
- The difference is in the probability of getting 2 heads in 3 coin flips

# Binomial Distribution

- Binomial = "two names", e.g., "heads" or "tails"
- Probability of two heads in three coin flips?
- $\Pr(\text{thh}) + \Pr(\text{hth}) + \Pr(\text{hht})$
- $q * p * p + p * q * p + p * p * q = 3 * p^2 * q^1$
- Multiply  $p$  for each head,  $q$  for each tail, then multiply by the number of combinations

$$\text{Prob}(K = k) = \binom{n}{k} * p^k * q^{(n-k)}$$

# Binomial Distribution Example

$$Prob(Y = k) = \binom{n}{k} * p^k * q^{(n-k)}$$

- $n = 3, k = 2, p = 0.9$
- $(n! / [(n - k)! * k!]) * p^k * (1 - p)^{(n - k)}$
- $(3! / [(3 - 2)! * 2!]) * 0.9^2 * 0.1^{(3 - 2)}$
- $(3! / [1! * 2!]) * 0.9^2 * 0.1$
- $3 * 0.9 * 0.9 * 0.1 = 0.243$  (or 24.3%)
- For a fair coin it would be 0.375 (or 37.5%)
- Biased coin, 3 heads:  $1 * 0.9^3 * 0.1^0 = 72.9\%$

# Flipping Coins Over Time

- Suppose we flip a very biased coin once a second
- If  $p = 1 / 60$ , we would expect to have one head and 59 tails each minute, on average
- In some minutes we might have more heads and in some minutes none at all
- To figure out the probability of getting a certain number of heads ( $k$ ) in a minute, we could use the binomial distribution ( $n = 60$ ,  $p = 1 / 60$ )
- We could add to figure out the cumulative probability of getting less than a number of heads in a minute, e.g., less than 3 heads:  $\Pr(k=0) + \Pr(k=1) + \Pr(k=2)$

# Events Per Second

- Now suppose we flip a coin every millisecond
- To keep the same event rate, use a constant
- Define  $\lambda$  (lambda) in units of heads per second
- $\lambda = p * n = \text{heads per flip} * \text{flips per second}$
- To increase  $n$  to 1,000 flips per second while keeping the event rate the same at  $\lambda = 1 / 60$  heads per second, the probability  $p$  would need to be adjusted to  $p = \lambda / n = ( 1 / 60 ) / 1,000 = 1 / 60,000$

# In the Limit, Part 1

$$\mathit{limit}_{n \rightarrow \infty} \left[ \binom{n}{k} \cdot p^k \cdot q^{(n-k)} \right]$$

$$\mathit{limit}_{n \rightarrow \infty} \left[ \left( \frac{n!}{(n-k)! \cdot k!} \right) \cdot \left( \frac{\lambda}{n} \right)^k \cdot \left( 1 - \frac{\lambda}{n} \right)^{(n-k)} \right]$$

$$\frac{\lambda^k}{k!} \cdot \mathit{limit}_{n \rightarrow \infty} \left[ \left( \frac{n!}{(n-k)!} \right) \cdot \left( \frac{1}{n} \right)^k \cdot \left( 1 - \frac{\lambda}{n} \right)^{(n-k)} \right]$$

## In the Limit, Part 2

$$\frac{\lambda^k}{k!} \cdot \lim_{n \rightarrow \infty} \left[ \left( \frac{n!}{(n-k)!} \right) \cdot \left( \frac{1}{n} \right)^k \cdot \left( 1 - \frac{\lambda}{n} \right)^{(n-k)} \right]$$

$$\frac{\lambda^k}{k!} \cdot \lim_{n \rightarrow \infty} \left[ \left( \frac{n!}{(n-k)!} \right) \cdot \left( \frac{1}{n} \right)^k \cdot \left( \frac{n}{n-\lambda} \right)^k \cdot \left( 1 - \frac{\lambda}{n} \right)^n \right]$$

$$\frac{\lambda^k}{k!} \cdot \lim_{n \rightarrow \infty} \left[ \left( \frac{n!}{(n-k)!} \right) \cdot \left( \frac{1}{n-\lambda} \right)^k \cdot \left( 1 - \frac{\lambda}{n} \right)^n \right]$$

# Pascal's Triangle

$$(x + y)^n = \sum_{k=0}^n \binom{n}{k} \cdot x^{(n-k)} \cdot y^k$$

1

$$1 \cdot x + 1 \cdot y$$

$$1 \cdot x^2 + 2 \cdot x \cdot y + 1 \cdot y^2$$

$$\binom{0}{0} \cdot x^{(0-0)} \cdot y^0$$

$$\binom{1}{0} \cdot x^{(1-0)} \cdot y^0 + \binom{1}{1} \cdot x^{(1-1)} \cdot y^1$$

$$\binom{2}{0} \cdot x^{(2-0)} \cdot y^0 + \binom{2}{1} \cdot x^{(2-1)} \cdot y^1 + \binom{2}{2} \cdot x^{(2-2)} \cdot y^2$$

# Pascal's Triangle Identity

$$(x + y)^n = \sum_{k=0}^n \binom{n}{k} \cdot x^{(n-k)} \cdot y^k$$

$$\left(1 + \frac{\alpha}{n}\right)^n = \sum_{k=0}^n \binom{n}{k} \cdot \left[\frac{\alpha}{n}\right]^k$$

# Exponential Function Identity

$$\lim_{n \rightarrow \infty} \left( 1 + \frac{\alpha}{n} \right)^n$$

$$\lim_{n \rightarrow \infty} \sum_{k=0}^n \binom{n}{k} \cdot \frac{\alpha^k}{n^k} = \lim_{n \rightarrow \infty} \sum_{k=0}^n \left[ \frac{n!}{k! \cdot (n-k)!} \right] \cdot \frac{\alpha^k}{n^k}$$

$$\lim_{n \rightarrow \infty} \sum_{k=0}^n \left[ \frac{n \cdot (n-1) \cdot (n-2) \cdot \dots \cdot (n-k+1)}{k!} \right] \cdot \frac{\alpha^k}{n^k}$$

$$\lim_{n \rightarrow \infty} \sum_{k=0}^n \left[ \frac{n \cdot (n-1) \cdot (n-2) \cdot \dots \cdot (n-k+1)}{n^k} \right] \cdot \frac{\alpha^k}{k!} = \sum_{k=0}^{\infty} \frac{\alpha^k}{k!} = e^{\alpha}$$

## In the Limit, Part 3

$$\frac{\lambda^k}{k!} \cdot \lim_{n \rightarrow \infty} \left[ \left( \frac{n!}{(n-k)!} \right) \cdot \left( \frac{1}{n-\lambda} \right)^k \cdot \left( 1 - \frac{\lambda}{n} \right)^n \right]$$

$$\lim_{n \rightarrow \infty} \left( 1 + \frac{-\lambda}{n} \right)^n = e^{-\lambda}$$

$$\frac{\lambda^k}{k!} \cdot e^{-\lambda} \cdot \lim_{n \rightarrow \infty} \left[ \frac{n \cdot (n-1) \cdot [\dots] \cdot (n-k+1)}{(n-\lambda)^k} \right]$$

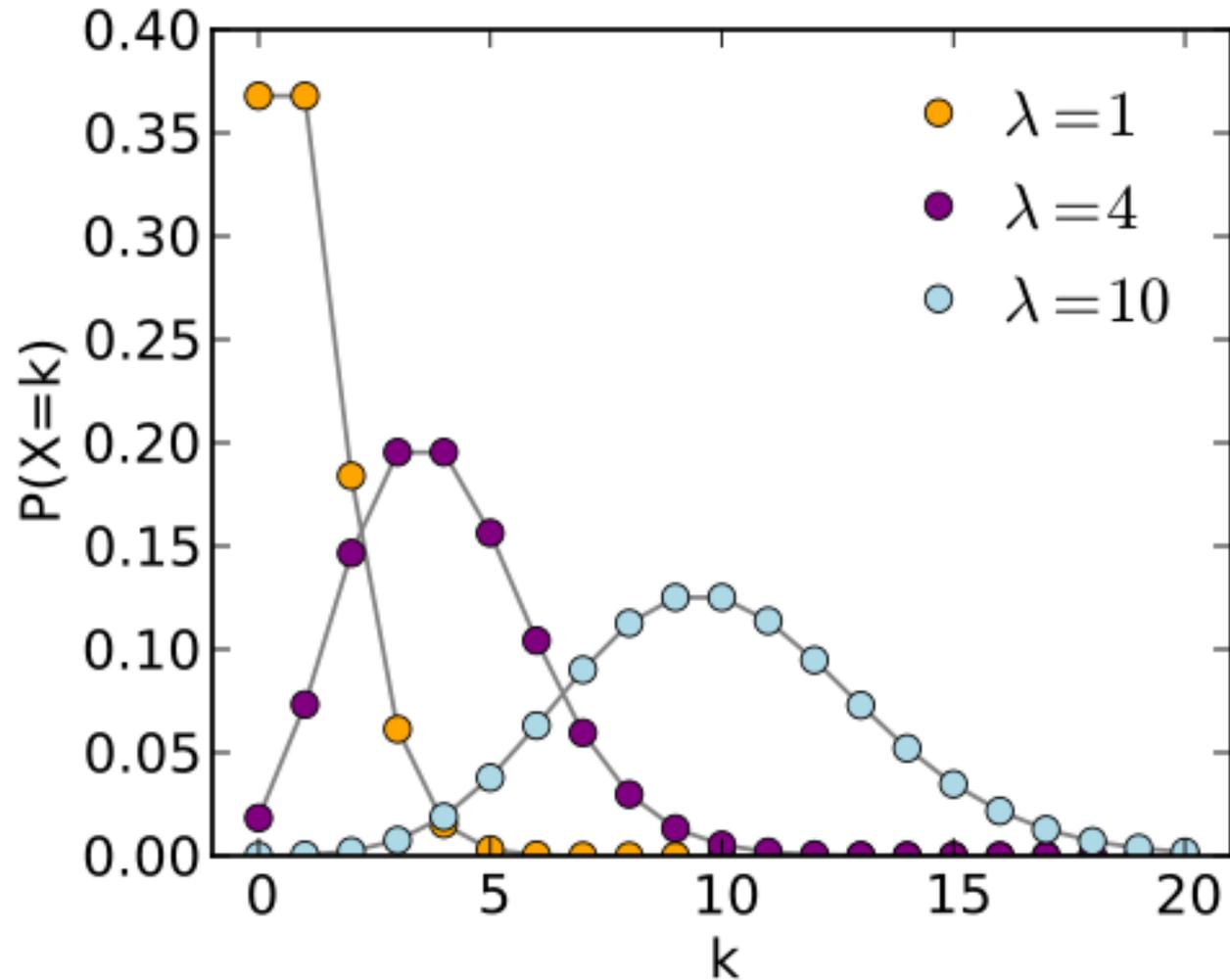
# In the Limit, Part 4

$$\frac{\lambda^k}{k!} \cdot e^{-\lambda} \cdot \lim_{n \rightarrow \infty} \left[ \frac{n \cdot (n-1) \cdot [\dots] \cdot (n-k+1)}{(n-\lambda)^k} \right]$$

$$\frac{\lambda^k}{k!} \cdot e^{-\lambda} \cdot \lim_{n \rightarrow \infty} \left[ \frac{n^k + [\dots]}{n^k + [\dots]} \right]$$

$$\frac{\lambda^k}{k!} \cdot e^{-\lambda}$$

# Poisson distribution



Skbkekak, "Plot of the probability mass function for the Poisson distribution",  
[http://en.wikipedia.org/wiki/File:Poisson\\_pmf.svg](http://en.wikipedia.org/wiki/File:Poisson_pmf.svg), 2010 Feb 10, viewed 2011 Jul 04,  
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# Poisson Distribution

$$Prob(K = k) = \frac{\lambda^k}{k!} \cdot e^{-\lambda}$$

- If  $\lambda$  is 20 events per second, what is the probability that there will be one or more events in a twentieth of a second?

$$Prob(K = k) = \frac{(\lambda \cdot \Delta T)^k}{k!} \cdot e^{-\lambda \cdot \Delta T}$$

- $\Delta T = 0.05$  seconds;  $\lambda * \Delta T = 1$
- $Prob(K \geq 1) = 1 - Prob(K = 0) = 1 - e^{-\lambda * \Delta T}$   
 $= 1 - e^{-1} = 1 - 0.367879 = 0.632121 = 63\%$

# Poisson Simulation

$$Prob(K = k) = \frac{(\lambda \cdot \Delta T)^k}{k!} \cdot e^{-\lambda \cdot \Delta T}$$

- If your computer simulation is being updated with a very small time step, what probability do you assign a Poisson distributed event?
- Assume two or more events in the same very small time step improbable
- At 20 events per second and  $\Delta T = 1$  ms,  
 $Prob(K \geq 1) = 1 - Prob(K = 0) = 1 - e^{-\lambda \cdot \Delta T}$   
 $= 1 - e^{-0.020} = 1 - 0.9801 = 0.0198 \approx 0.020$
- NB:  $Prob(K \geq 1) \approx \lambda \cdot \Delta T$  for small  $\lambda \cdot \Delta T$

# Poisson Process

- Events occur continuously at some fixed rate
  - But are assumed not to occur simultaneously
- Events occur independently of each other
  - The property of memorylessness
- Examples: the count for a given unit of time of
  - Atoms that radioactively decay
  - Raindrops that land on a plate
  - Visitors to a website
  - Neural action potentials

# Mean = Rate

$$E(K) = \sum_{k=0}^{\infty} k \cdot p(k, \lambda) = \sum_{k=0}^{\infty} k \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right]$$

$$\lambda \cdot e^{-\lambda} \cdot \sum_{k=1}^{\infty} \frac{\lambda^{(k-1)}}{(k-1)!}$$

$$\lambda \cdot e^{-\lambda} \cdot \sum_{j=0}^{\infty} \frac{\lambda^j}{j!} \text{ where } j = k - 1$$

$$\lambda \cdot e^{-\lambda} \cdot e^{\lambda} = \lambda$$

# Variance = Mean, Part 1

$$\text{Var}(K) = \sum_{k=0}^{\infty} (k - \lambda)^2 \cdot p(k, \lambda) = \sum_{k=0}^{\infty} (k^2 - 2 \cdot k \cdot \lambda + \lambda^2) \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right]$$

$$\sum_{k=0}^{\infty} k^2 \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] - 2 \cdot \lambda \cdot \sum_{k=0}^{\infty} k \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] + \lambda^2$$

$$\sum_{k=0}^{\infty} k^2 \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] - 2 \cdot \lambda \cdot \sum_{k=0}^{\infty} k \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] + \lambda^2 \sum_{k=0}^{\infty} \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right]$$

$$\sum_{k=0}^{\infty} k^2 \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] - 2 \cdot \lambda^2 + \lambda^2 \cdot e^{-\lambda} \cdot e^{\lambda}$$

$$\sum_{k=0}^{\infty} k^2 \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] - \lambda^2$$

# Variance = Mean, Part 2

$$\sum_{k=0}^{\infty} k^2 \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] - \lambda^2 = \sum_{k=0}^{\infty} [k \cdot (k-1) + k] \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] - \lambda^2$$

$$\sum_{k=0}^{\infty} [k \cdot (k-1)] \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] + \sum_{k=0}^{\infty} k \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] - \lambda^2$$

$$\sum_{k=0}^{\infty} [k \cdot (k-1)] \cdot \left[ \frac{\lambda^k}{k!} \cdot e^{-\lambda} \right] + \lambda - \lambda^2 = e^{-\lambda} \cdot \sum_{k=2}^{\infty} [k \cdot (k-1)] \cdot \left[ \frac{\lambda^k}{k!} \right] + \lambda - \lambda^2$$

$$e^{-\lambda} \cdot \lambda^2 \cdot \sum_{k=2}^{\infty} \left[ \frac{\lambda^{(k-2)}}{(k-2)!} \right] + \lambda - \lambda^2 = e^{-\lambda} \cdot \lambda^2 \cdot e^{\lambda} + \lambda - \lambda^2 = \lambda$$

# References

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