## **Gamma Function**

1) 
$$\Gamma(t) = \int_{0}^{\infty} y^{t-1} e^{-y} dy \quad \Rightarrow \text{ integration by parts } (t-1) \text{ times}$$
  
2)  $\Gamma(1) = \int_{0}^{\infty} y^{1-1} e^{-y} dy = \int_{0}^{\infty} e^{-y} dy = 1$   
3)  $\Gamma(t) = y^{t-1} (-e^{-y}) \Big|_{0}^{\infty} - \int_{0}^{\infty} (-e^{-y})(t-1) y^{t-2} dy$   
 $= (0 - 0) + (t-1) \int_{0}^{\infty} y^{t-2} e^{-y} dy$ 

## **<u>Gamma RVs</u>** $X \sim Gamma(\alpha, \theta)$

- used in a variety of modeling situations, including waiting times in a Poisson Process
- extends the Exponential RV (similar to the Negative Binomial vs. Geometric RV)
- $\circ \alpha$  is a shape parameter (represents the number of successes in a Poisson Process)
- $\theta$  is a scale parameter (represents the mean waiting time between successes,  $\theta = 1/\lambda$ )

• PDF: 
$$f(x) = \begin{cases} \frac{1}{\Gamma(\alpha)\theta^{\alpha}} x^{\alpha-1} e^{-\frac{x}{\theta}} & x > 0\\ 0 & \text{otherwise} \end{cases}$$

• MGF: 
$$M(t) = \int_{-\infty}^{\infty} e^{tx} f(x) dx = \frac{1}{(1-\theta t)^{\alpha}}$$

$$\circ \ \alpha = l \quad \Rightarrow \quad \frac{1}{\Gamma(1)\theta^{1}} x^{1-1} e^{-\frac{x}{\theta}} = \frac{1}{\theta} e^{-\frac{x}{\theta}} \quad \Rightarrow \text{ Exponential PDF}$$

## Poisson Process (2)

- A large number of trials with a small *P*(*success*) on any trial.
- # Successes in a given time interval is a Discrete RV and is modeled by a Poisson RV.
- $\circ \lambda$  is the mean # of successes per unit time ( $\lambda$  = mean rate)
- $\circ$  ( $\lambda t$ ) is the mean # of successes in an interval of length t
- The time until the <u>first</u> successes is a continuous RV and is modeled by an <u>Exponential RV</u>.
- The time until the  $\underline{\alpha}^{th}$  success is a continuous RV and is modeled by a <u>Gamma RV</u>.

Let **T**= waiting time until there are  $\alpha$  successes

 $F(t) = P(T \le t) = 1 - P(T > t)$ 

=1-P(less than  $\alpha$  Successes in [0,t])

## **Chi-Square Goodness-of-fit Tests**

- Used to check if data comes from a specific distribution (Probability Model)
- Compares observed and expected counts under the Probability Model in each of k categories

$$\chi^{2}_{GOF} = \sum_{i=1}^{k} \frac{(O_{i} - E_{i})^{2}}{E_{i}}$$

• When the Probability Model is correct,  $\chi^2_{GOF}$  follows a  $\chi^2_{(r)}$  distribution with r=k-1-e

- <u>Genetic Models</u>
- Hardy-Weinberg Model (independence, natural selection, genotyping errors, ...)
- Dominant/Recessive Models (Punnett Squares)
  - A Hybrid cross  $(Aa \times Aa)$

	A	а
A	AA	Aa
a	aA	aa

- If *A* is dominant to *a* then we expect offspring in a ratio of 3:1
- Example: For a certain plant *Purple* flowers are thought to be dominant to *White* 
  - A self-pollination study of 400 plants from a Hybrid cross gives 278 *Purple &* 122 *White*
  - 1. Find the Expected numbers of *Purple & White* under the dominance model.
  - 2. Determine the  $X^2_{GOF}$  value for these data.
  - 3. Find the Probability Value (p-value) from the Chi-square table.