

## Handout 4: THE POISSON PROCESS

### Math/Stat 394: Probability I

Wellner; 1/24/2000

$N(t) \equiv$  (the total *count* at time  $t$ )  $\sim \text{Poisson}(\nu t)$   
 $\nu$  is the *intensity*, or mean number of counts per unit of time  
 $\theta \equiv 1/\nu$  will be seen to be the mean time between counts.  
Examples: traffic accidents, telephone calls, defects per foot, ...  
 $Y_i \equiv$  ( the  $i$ -th interarrival time )  $\sim \text{Exponential}(\nu)$   
 $f_{Y_i}(t) = \nu \exp(-\nu t) 1_{(0,\infty)}(t)$  with  
 $E(Y_i) = 1/\nu = \theta$  and  $\text{Var}(Y) = 1/\nu^2 = \theta^2$ .

$W_r \equiv Y_1 + \cdots + Y_r$  ( the waiting time until the  $r$ -th event )  
 $\sim \text{Gamma}(r, \nu)$ ;  
 $f_{W_r}(t) = (\nu t)^{r-1} / \Gamma(r) \nu \exp(-\nu t) 1_{(0,\infty)}(t)$ ,  
with  $E(W_r) = r/\nu$  and  $\text{Var}(W_r) = r/\nu^2$ .

- Key facts:  $[Y_1 > t] = [N(t) = 0]$  and  $[W_r > t] = [N(t) < r]$ .

- Poisson facts:

(a) The rv  $N(t) \sim \text{Poisson}(\nu t)$ .  
 $E(N(t)) = \nu t$  and  $\text{Var}(N(t)) = \nu t$ .

(b) The rv

$$Z_t \equiv \frac{N(t) - \mu_t}{\sigma_t} = \frac{N(t) - \nu t}{\sqrt{\nu t}}$$

is approximately  $N(0, 1)$  for large values of  $\nu t$ .

(c)  $N(s) \sim \text{Poisson}(\nu s)$  and  $N(t) - N(s) \sim \text{Poisson}(\nu(t - s))$  are independent, and their sum,  $N(t) \sim \text{Poisson}(\nu t)$ .

(e) Given that  $N(t) = m$ , for an integer  $m \geq 1$ , for  $0 < s < t$   
 $(N(s) | N(t) = m) \sim \text{Binomial}(m, s/t)$ .

- Gamma facts:

(a) The first waiting time rv  $Y \equiv Y_1$  has mean  $E(Y) = 1/\nu = \theta$  and variance  $\text{Var}(Y) = 1/\nu^2 = \theta^2$ .

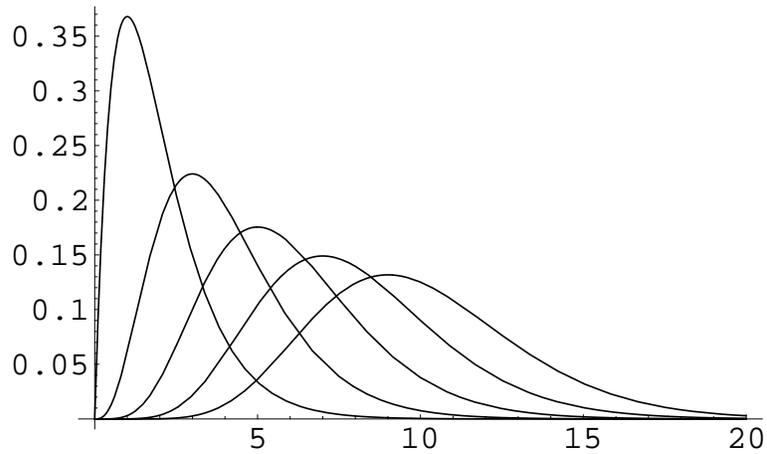


Figure 1: Plot of Gamma( $k, 1$ ) densities  $k = 2, 4, 6, 8, 10$ .

(b) The rv  $W_r$  has mean  $E(W_r) = r/\nu = r\theta$  and variance  $Var(W_r) = r/\nu^2 = r\theta^2$ .

(c) The rv

$$Z_r \equiv \frac{W_r - E(W_r)}{\sigma_r} = \frac{W_r - r\theta}{\sqrt{r\theta^2}}$$

is approximately  $N(0, 1)$ .

(d) If  $W_r \sim \text{Gamma}(r, \nu)$  and  $W_s \sim \text{Gamma}(s, \nu)$  are independent, then  $W_r + W_s \sim \text{Gamma}(r + s, \nu)$ .

- Exponential facts:

(a) If  $Y \equiv Y_1$ ,  $P(Y > t) = P(N(t) = 0) = e^{-\nu t}$  for  $t \geq 0$ .

Hence  $f_Y(t) = (d/dt)F_Y(t) = (d/dt)(1 - e^{-\nu t}) = \nu e^{-\nu t}$  for  $t \geq 0$ .

(b)  $P(Y > t + s | Y > s) = P(Y > t + s) / P(Y > s) = e^{-\nu(t+s)} / e^{-\nu s} = e^{-\nu t} = P(Y > t)$ . This is the *memoryless* property of the Exponential distribution.

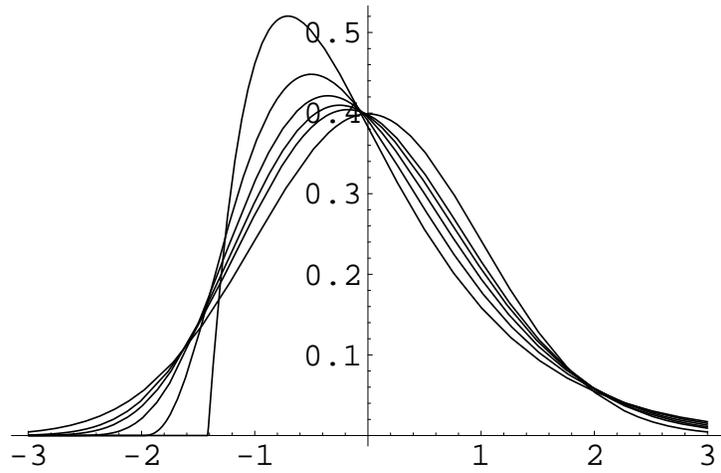


Figure 2: Plots of standardized  $\text{Gamma}(k, 1)$  densities  $k = 2, 4, 8, 16, 32$ .

- Conditional on either  $N(t_0) = m$  or  $W_{m+1} = t_0$ , there are  $m$  arrivals uniformly distributed over  $[0, t_0]$ . That is, given the number  $m$  of events on or before a given time, the location is random and does not depend on  $\nu$ .