

# A Concise Summary

## Everything you need to know about exponential and Poisson

### Exponential Distribution

Assume that  $X \sim \exp(\lambda)$ , by which we mean that  $X$  has an exponential distribution with rate  $\lambda$ . Then  $X$  has mean  $1/\lambda$ ; i.e.,  $EX = 1/\lambda$ . Also the variance is  $Var(X) = (EX)^2 = 1/\lambda^2$ . In addition, assume that  $Y \sim \exp(\mu)$  and  $X_i \sim \exp(\lambda_i)$  for  $i = 1, \dots, n$ , where all these exponential random variables are independent.

1. *Lack of memory*:  $P(X > s + t | X > s) = P(X > t)$  for all  $s > 0$  and  $t > 0$ .  
(check the computation)
2. *Minimum*:  $\min\{X, Y\} \sim \exp(\lambda + \mu)$   
(check the computation)  
and hence  
 $\min\{X_1, \dots, X_n\} \sim \exp(\lambda_1 + \dots + \lambda_n)$  without computation.
3. *Maximum*:  $X + Y = \min\{X, Y\} + \max\{X, Y\}$  tells us an easy way to compute  $E[\max\{X, Y\}]$ .
4. *More on Minimum*:  $P(X = \min\{X, Y\}) = P(X < Y) = \frac{\lambda}{\lambda + \mu}$  : (check the computation)  
and hence  
 $P(X_k = \min\{X_1, \dots, X_n\}) = \frac{\lambda_k}{\lambda_1 + \dots + \lambda_n}$  without computation.
5. *Even more on Minimum*: The events  $\{X = \min\{X, Y\}\}$  and  $\{\min\{X, Y\} > t\}$  are independent for all  $t$ .  
and hence  
the events  $\{X_k = \min\{X_1, \dots, X_n\}\}$  and  $\{\min\{X_1, \dots, X_n\} > t\}$  are independent for all  $t$ .

### Poisson Processes

Consider a Poisson process  $\{N(t) : t \geq 0\}$  with rate  $\lambda$ , referred to by  $N(t)(\lambda)$ . In addition, consider Poisson processes  $N_j(t)(\lambda_j)$ ,  $1 \leq j \leq m$ .

1. *Interarrival Times*: The interarrival times of  $N(t)(\lambda)$  are IID  $\exp(\lambda)$  .
2. *Thinning (Type classification)* : When arrivals occur in the Poisson process  $N(t)$ , they are classified randomly and *independently* (the successive classifications are done independently according to the same probabilities) into classes (indexed by  $j$ ) with probability  $p_1, \dots, p_m$ . Let  $N_j(t)$  be the input (arrival) process for class  $j$  (obtained with probability  $p_j$ ). These newly created counting processes  $N_j(t)$  are independent Poisson processes with rates  $\lambda p_j$ .
3. *Superposition (Type aggregation)* : When *independent* Poisson arrivals  $N_j(t)$  occur with rates  $\lambda_j$ , the total number of arrivals is a Poisson( $\lambda_1 + \dots + \lambda_m$ ).
4. *Conditioning* : When we know  $N(t) = n$ , the occurrence time of  $n$  arrivals are distributed as independent random variables, each uniform on the interval  $[0, t]$ .