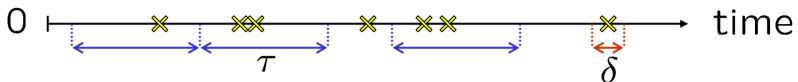
#### **LECTURE 18**

Readings: Finish Section 5.2

### Lecture outline

- Review of the Poisson process
- Properties
  - Adding Poisson Processes
  - Splitting Poisson Processes
- Examples

#### The Poisson Process: Review



• Number of arrivals in disjoint time intervals are independent,  $\lambda$  = "arrival rate"

$$\mathbf{P}(k,\delta) pprox egin{cases} 1 - \lambda \delta & ext{if } k = 0 \ \lambda \delta & ext{if } k = 1 \ 0 & ext{if } k > 0 \end{cases}$$
 (for very small  $\delta$ )
 $\mathbf{P}(k,\tau) = \frac{(\lambda t)^k}{k!} e^{-\lambda t} \qquad \mathbf{E}(N) = \lambda \tau$  (Poisson)

• Interarrival times (k = 1):

$$f_{Y_1}(y) = \lambda e^{-\lambda y}, \quad y \ge 0$$
 (Exponential)

• Time to the  $k^{th}$  arrival:

$$f_{Y_k}(y) = \frac{\lambda^k y^{k-1} e^{-\lambda y}}{(k-1)!}, \quad y \ge 0$$
 (Erlang)

#### Example: Poisson Catches

- Catching fish according to Poisson  $\lambda = 0.6/\text{hour}$  .
- Fish for two hours, but if there's no catch, continue until the first one.
  - ightharpoonup P(fish more than 2 hrs) =
  - ightharpoonup P(fish more than 2 but less than 5 hrs) =

ightharpoonup P(catch at least 2 fish) =

### Example: **Poisson Catches**

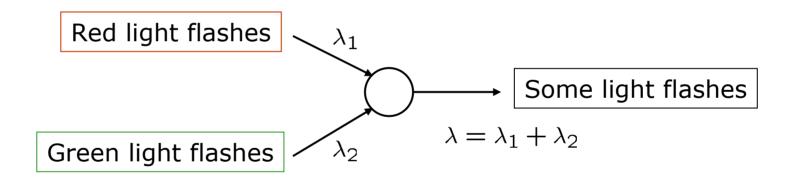
- Catching fish according to Poisson  $\lambda = 0.6/\text{hour}$  .
- Fish for two hours, but if there's no catch, continue until the first one.
  - ightharpoonup E[number of fish] =

 $\rightarrow$  E[future fishing time | fished for 4 hrs] =

ightharpoonup E[total fishing time] =

### **Adding (Merging) Poisson Processes**

- Sum of independent Poisson random variables is Poisson.
- Sum of independent Poisson processes is Poisson.

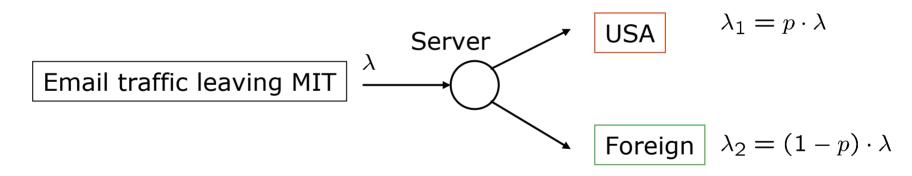


 What is the probability that the next arrival comes from the first process?

$$\frac{\lambda_1 \delta}{\lambda_1 \delta + \lambda_2 \delta} = \frac{\lambda_1}{\lambda_1 + \lambda_2}$$

### **Splitting of Poisson Processes**

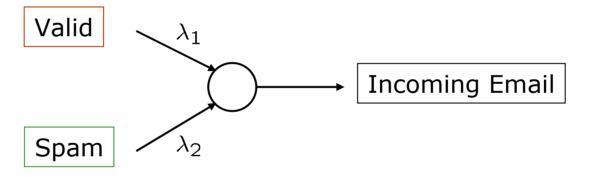
- ullet Each message is routed along the first stream with probability p , and along the second stream with probability 1-p .
  - Routing of different messages are independent.



- Each output stream is Poisson.

### Example: **Email Filter** (1)

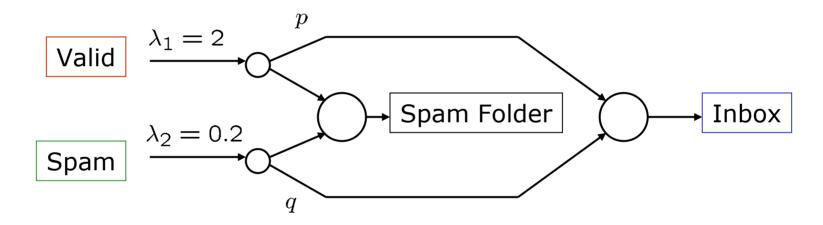
- You have incoming email from two sources: valid email, and spam. We assume both to be Poisson.
- Your receive, on average, 2 valid emails per hour, and 1 spam email every 5 hours.



- Total incoming email rate =  $\lambda = \lambda_1 + \lambda_2 = 2.2$  emails per hour.
- Probability that a received email is spam =  $\frac{\lambda_2}{\lambda_1 + \lambda_2} = \frac{0.2}{2.2} \approx 0.09$

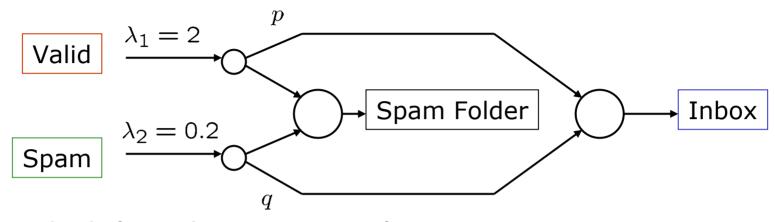
## Example: Email Filter (2)

 You install a spam filter, that filters out spam email correctly 80% of the time, but also identifies a valid email as spam 5% of the time.



- p = 0.95 q = 0.2
- Inbox email rate =  $p\lambda_1 + q\lambda_2 = 0.95 \cdot 2 + 0.2 \cdot 0.2 = 1.94$
- Spam folder email rate = 2.2 1.94 = 0.26

# Example: Email Filter (3)



- Probability that an email in the inbox is spam =  $\frac{q\lambda_2}{p\lambda_1+q\lambda_2}=\frac{0.2\cdot0.2}{1.94}\approx0.02$
- Probability that an email in the spam folder is valid =  $\frac{(1-p)\lambda_1}{(1-p)\lambda_1+(1-q)\lambda_2} = \frac{0.05\cdot 2}{0.26} \approx 0.38$
- Every how often should you check your spam folder, to find one valid email, on average?

$$E(N) = \lambda_1 (1 - p)\tau = 1 \Rightarrow \tau = \frac{1}{0.05 \cdot 2} = 10$$
 hrs.