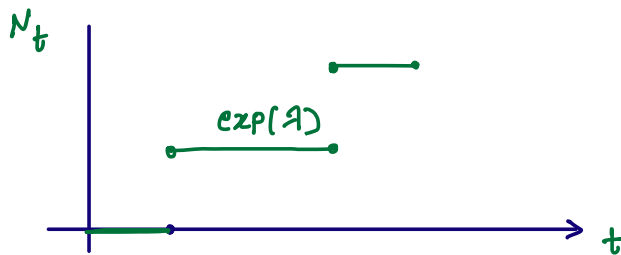


Poisson process: N_t is a counting process with rate λ

① $N_t - N_s \perp \{N_r\}_{r \leq s} \rightarrow$ indep increment

② $N_t - N_s \sim \text{pois}(\lambda(t-s)) \quad \forall t \geq s \rightarrow$ stationary



super position

$$X \sim \text{pois}(\lambda)$$

$$X \perp Y$$

$$Y \sim \text{Pois}(\gamma)$$

1st approach

$$\begin{aligned} \Pr(X+Y=k) &= \Pr(X=0, Y=k) + \Pr(X=1, Y=k-1) + \dots \\ &= \Pr(X=0) \times \Pr(Y=k) + \Pr(X=1) \times \Pr(Y=k-1) + \dots \end{aligned}$$

2nd approach

N_t is a poisson process with rate γ .

$$N_\mu \sim \text{pois}(\mu) \sim X$$

$$N_{\gamma+\mu} - N_\mu \sim \text{Pois}(\gamma) \sim Y$$

$X+Y$ has the same distribution as $(N_{\gamma+\mu} - N_\mu) + N_\mu$

which is $\text{Pois}(\tau + \nu)$

superposition is the generalization of the above idea

Let us define

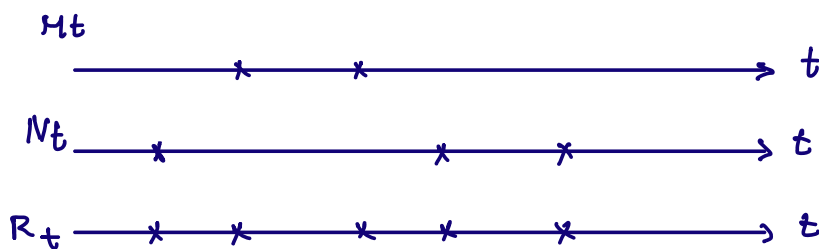
$M_t = \#$ undergrads that arrive by time t

$N_t = \#$ grads that arrive by time t

$R_t = \#$ customers that arrive by time t

$\{M_t\}_{t \geq 0}$ is a poisson process with rate ν

$\{N_t\}_{t \geq 0}$ " " " " " " τ



$$R_t - R_s = (M_t - M_s) + (N_t - N_s) \stackrel{?}{\perp\!\!\!\perp} \begin{matrix} \{R_u\}_{u \leq s} \\ \downarrow \\ \{M_u + N_u\}_{u \leq s} \end{matrix}$$

$$M_t - M_s \perp\!\!\!\perp \{M_u\}_{u \leq s}$$

$$M_t - M_s \perp\!\!\!\perp \{N_u\}_{u \leq s}$$

exactly the same reasoning applies to $N_t - N_s$

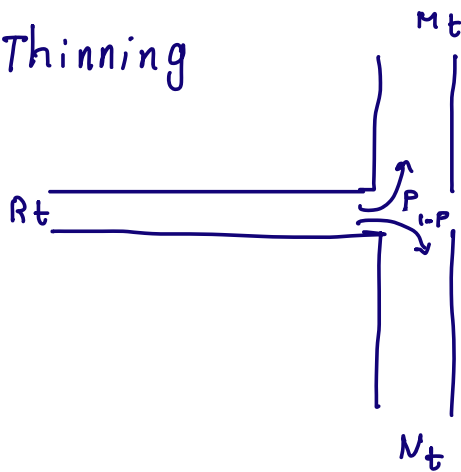
$$M_t - M_s \sim \text{pois}(\mu(t-s)) \quad \Rightarrow \quad R_t - R_s \sim \text{pois}((\mu+r)(t-s))$$

$$N_t - N_s \sim \text{pois}(r(t-s))$$

$\Rightarrow R_t$ is a poisson process with rate $(\mu+r)$

If $N_t^1, N_t^2, \dots, N_t^k$ are indep poisson processes with rates $\mu_1, \mu_2, \dots, \mu_k$, then $N_t^1 + \dots + N_t^k$ is a poisson process with rate $\mu_1 + \dots + \mu_k$

Thinning



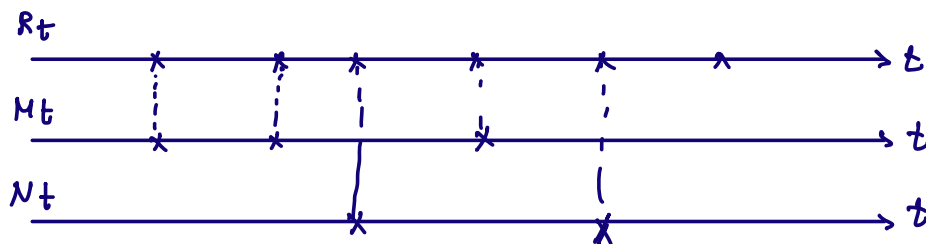
$R_t = \#$ customers arrived by time t

$M_t = \#$ customers going Up

$N_t = \#$ customers going Down

$X_k =$ which door customer k took

$$P\{X_k = U\} = p \quad \& \quad P(X_k = D) = (1-p)$$



$$M_t = \sum_{k=1}^{R_t} \mathbb{1}\{X_k = U\}$$

$$N_t = \sum_{k=1}^{\infty} \mathbb{1}_{\{X_k = D\}}$$

$$\begin{aligned} P\{M_t = k\} &= \sum_{r=0}^{\infty} P\{M_t = k | R_t = r\} P(R_t = r) \\ &= \sum_{r=0}^{\infty} P\left\{ \sum_{i=1}^{R_t} \mathbb{1}_{\{X_i = U\}} = k \mid R_t = r \right\} \times P(R_t = r) \\ &= \sum_{r=0}^{\infty} P\left\{ \sum_{i=1}^r \mathbb{1}_{\{X_i = U\}} = k \mid R_t = r \right\} P(R_t = r) \\ &= \sum_{r=0}^{\infty} P\left\{ \sum_{i=1}^r \mathbb{1}_{\{X_i = U\}} = k \right\} P(R_t = r) \end{aligned}$$

$$\begin{aligned} P\{M_t = k\} &= \sum_{r=k}^{\infty} \binom{r}{k} p^k (1-p)^{r-k} \frac{e^{-\mu t} (\mu t)^r}{r!} \\ &= \sum_{r=k}^{\infty} \frac{r!}{k! (r-k)!} p^k (1-p)^{r-k} e^{-\mu t} \frac{(\mu t)^r}{r!} \\ &= \frac{e^{-\mu t} (p \mu t)^k}{k!} \sum_{r=k}^{\infty} \frac{((1-p) \mu t)^{r-k}}{(r-k)!} \\ &= \frac{e^{-\mu t} (p \mu t)^k}{k!} \times e^{(1-p) \mu t} \\ &= \frac{e^{-\mu p t} (\mu p t)^k}{k!} \end{aligned}$$

It is easy to show that M_t has indep increment

M_t is a poisson process with rate $p \cdot \nu$

$N_t \sim \dots \sim (1-p) \nu$

Note that

$$M_t + N_t = R_t$$

$$P(M_t = m, N_t = n) = P(M_t = m, R_t = m+n)$$

$$= P\left(\sum_{i=1}^{R_t} \mathbb{1}_{\{X_i \leq U\}} = m, R_t = m+n\right)$$

$$= P\left(\sum_{i=1}^{m+n} \mathbb{1}_{\{X_i \leq U\}} = m, R_t = m+n\right)$$

$$= P\left(\sum_{i=1}^{m+n} \mathbb{1}_{\{X_i \leq U\}} = m\right) P(R_t = m+n)$$

$$= \binom{m+n}{m} p^m (1-p)^n e^{-\mu t} \frac{(\mu t)^{m+n}}{(m+n)!}$$

$$= \frac{e^{-\mu p t} (p \mu t)^m}{m!} \times \frac{e^{-(1-p)\mu t} ((1-p)\mu t)^n}{n!}$$

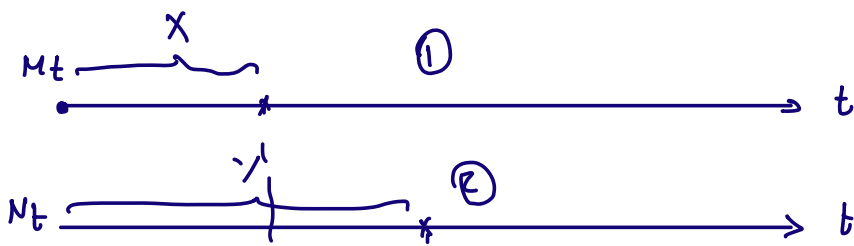
$$= P(M_t = m) \times P(N_t = n)$$


Ex: Let $X \sim \text{exp}(\mu)$ & $Y \sim \text{exp}(\gamma)$

$$X \wedge Y$$

Let M_t be a poisson process with rate μ

" N_t " " " " " " γ



$$R_t = M_t + N_t$$


for thinning of $R_t \longrightarrow M_t$ we need to
 set the probability to $\frac{\mu}{\mu+r}$

$$R_t \longrightarrow N_t \quad \text{subsampling } \frac{r}{\mu+r}$$

$$P(X \leq Y) = P(\underline{X \wedge Y} = X) = P(\text{first label is } \textcircled{1}) \\ = \frac{\mu}{\mu+r}$$

$X \wedge Y$ is the first arrival of $R_t \sim \exp(\mu+r)$