

Computing expectations:Cont. case: $\bar{X} \geq 0$, 1st moment

$$\begin{aligned}
 E[\bar{X}] &= \int_0^{\infty} x \varphi(x) dx & F(x) &= \int_0^x \varphi(x) dx, \quad F'(x) = \varphi(x) \\
 & & \text{Integrate by parts: } & u = x, \quad v' = \varphi(x) \\
 & & & u' = 1, \quad v = F(x) - 1 \\
 &= \lim_{c \rightarrow \infty} \int_0^c x \varphi(x) dx = \lim_{c \rightarrow \infty} \left([x(F(x)-1)]_0^c - \int_0^c (F(x)-1) dx \right) \\
 &= \lim_{c \rightarrow \infty} \left(\underbrace{c(F(c)-1)}_{\substack{\uparrow \\ \text{1st} \\ \text{moment}}} + \int_0^c \underbrace{[1-F(x)]}_{P[\bar{X} > x]} dx \right) = \int_0^{\infty} P[\bar{X} > x] dx
 \end{aligned}$$

Discrete case: \bar{X} taking values in $\{0, 1, 2, 3, \dots\}$

$$\text{1 moment exists: } E[\bar{X}] = \sum_{n=1}^{\infty} P[\bar{X} \geq n].$$

Example (collecting coupons, football cards)

N football players: Prob of getting player i is p_i .

How long does it take to get a complete collection?

We buy cards according to a Poisson process of rate λ : N .

N_i is the process you get by just counting cards with player i :

N_i is Poisson with rate $p_i \lambda$, and the N_i 's are independent.

Let Σ_i be the first arrival time of N_i .

$$\Sigma = \max_i \{\Sigma_i\}$$

$$P[\Sigma \leq t] = P[\Sigma_1 \leq t \wedge \Sigma_2 \leq t \wedge \dots \wedge \Sigma_n \leq t]$$

$$\stackrel{\text{independence}}{=} P[\Sigma_1 \leq t] P[\Sigma_2 \leq t] \dots P[\Sigma_n \leq t] \text{ (exponentially dist)}$$

$$= (1 - e^{-\lambda p_1 t}) (1 - e^{-\lambda p_2 t}) \dots (1 - e^{-\lambda p_n t})$$

$$= \prod_{i=1}^n (1 - e^{-\lambda p_i t})$$

$$\text{Observe that } P[\Sigma > t] = 1 - \prod_{i=1}^n (1 - e^{-\lambda p_i t})$$

Hence:

$$E[\Sigma] = \int_0^{\infty} P[\Sigma > t] dt = \int_0^{\infty} (1 - \prod_{i=1}^n (1 - e^{-\lambda p_i t})) dt.$$

What about the number of cards we have to buy:

$\lambda = 1$: Let T_1, T_2, \dots be the interarrival times of Σ .

$$\Sigma = T_1 + T_2 + \dots + T_{\textcircled{N}} \leftarrow \text{when } \textcircled{N} \text{ has a complete collection}$$

$$= \sum_{i=1}^{\infty} \mathbb{1}_{\{i \leq N\}} T_i$$

↑ is 1 when $i \leq N$ and 0 when $i > N$

Hence

$$E[\Sigma] = E\left[\sum_{i=1}^{\infty} \mathbb{1}_{\{i \leq N\}} T_i\right] \stackrel{\text{independence}}{=} \sum_{i=1}^{\infty} \underbrace{E[\mathbb{1}_{\{i \leq N\}}]}_{P[N \geq i]} \underbrace{E[T_i]}_{1}$$

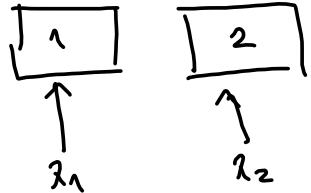
$$= \sum_{i=1}^{\infty} P[N \geq i] = E[N]$$

Thus

$$E[N] \stackrel{\lambda=1}{=} E[\Sigma] = \int_0^{\infty} (1 - \prod_{i=1}^n (1 - e^{-p_i t})) dt$$

Example: $N=23$ $p_i = \frac{1}{23}$ $E[N] = 8.6$.

New problem: $S_n^{(1)}$ = quest no n at 1



$S_m^{(2)}$ = quest no m at 2

$P[S_n^{(1)} \leq S_m^{(2)}]$

$n = m = 1$: $P[S_1^{(1)} \leq S_1^{(2)}] = \frac{\lambda_1}{\lambda_1 + \lambda_2}$

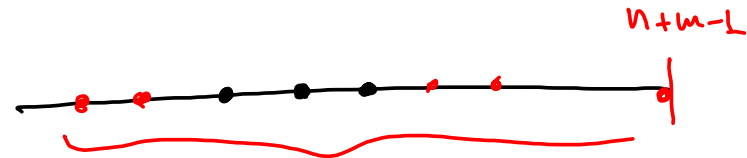
$n = 2, m = 1$ $P[S_2^{(1)} \leq S_1^{(2)}] = \frac{\lambda_1}{\lambda_1 + \lambda_2} \cdot \frac{\lambda_1}{\lambda_1 + \lambda_2} = \left(\frac{\lambda_1}{\lambda_1 + \lambda_2}\right)^2$

$P[S_n^{(1)} \leq S_m^{(2)}]$

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



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



$\sum_{k=n}^{n+m-1} \binom{n+m-1}{k} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2}\right)^k \left(\frac{\lambda_2}{\lambda_1 + \lambda_2}\right)^{n+m-1-k}$