

Poisson paradigm *Binomial case: $np \approx \lambda$
Here: $p_i \approx \frac{\lambda}{n}$*

Situation: Consider

- n events E_1, \dots, E_n *n should be large*
- $p_i = \mathbf{P}(E_i) \rightarrow p_i$ *should be small* *or* $\mathbf{P}(E_i | E_j) \approx \frac{1}{n}$
- Weak dependence of the E_i : $\mathbf{P}(E_i E_j) \approx \frac{1}{n}$
- $\lim_{n \rightarrow \infty} \sum_{i=1}^n p_i = \lambda$

Heuristic limit: Under the conditions above we expect that

times that E_i occurs $= X_n = \sum_{i=1}^n \mathbf{1}_{E_i} \rightarrow \mathcal{P}(\lambda)$ (3)

$= \begin{cases} 1 & \text{if } E_i \text{ occurs} \\ 0 & \text{otherwise} \end{cases}$

Example: matching problem (1)

Situation:

- n men take off their hats
- Hats are mixed up
- Then each man selects his hat at random
- Match: if a man selects his own hat

Question: Compute

- $\mathbf{P}(E_k)$ with $E_k =$ "exactly k matches"

Random variable . Let

$X = \# \text{ matches}$

Then exactly k matches

$$P(\overbrace{E_k}^{\uparrow}) = P(X = k)$$

Decomposition for X

$G_i \equiv$ "person i gets his own hat"

$$\Rightarrow X = \sum_{i=1}^n \mathbb{1}_{G_i}$$

Summary $X = \sum \mathbb{1}_{G_i}$, $P(E_k) = P(X=k)$

We wish to prove $\sum_{i=1}^n \frac{1}{i} \rightarrow \infty$

(i) $p_i = P(G_i) \approx \frac{1}{n}$

(ii) $P(G_j | G_i) \approx \frac{1}{n}$ if $j \neq i$

Here

(i) $P(G_i) = P(\text{person } i \text{ picks his own hat among } n \text{ hats})$

$$= \frac{1}{n} = p_i \Rightarrow \sum_{i=1}^n p_i = \sum_{i=1}^n \frac{1}{n} = 1 \equiv \Delta$$

Example: matching problem (2)

Fact: Using heavy combinatorics, one can prove

$$\mathbf{P}(E_k) = \frac{1}{k!} \sum_{j=2}^{n-k} \frac{(-1)^j}{j!}$$

Thus

$$\lim_{n \rightarrow \infty} \mathbf{P}(E_k) = \frac{e^{-1}}{k!}$$

New events: We set

$G_i =$ "Person i selects his own hat"

Example: matching problem (3)

Probabilities for G_j : We have

$$\mathbf{P}(G_i) = \frac{1}{n}, \quad \mathbf{P}(G_i | G_j) = \frac{1}{n-1}$$

Random variable of interest:

$$X = \sum_{i=1}^n \mathbf{1}_{G_i} \implies \mathbf{P}(E_k) = \mathbf{P}(X = k)$$

Poisson paradigm: From (3) we have $X \simeq \mathcal{P}(1)$. Therefore

$$\mathbf{P}(E_k) = \mathbf{P}(X = k) \simeq \mathbf{P}(\mathcal{P}(1) = k) = \frac{e^{-1}}{k!}$$

Outline

- 1 Random variables
- 2 Discrete random variables
- 3 Expected value
- 4 Expectation of a function of a random variable
- 5 Variance
- 6 The Bernoulli and binomial random variables
- 7 The Poisson random variable
- 8 Other discrete random variables**
- 9 Expected value of sums of random variables
- 10 Properties of the cumulative distribution function

Geometric random variable

Notation:

$$X \sim \mathcal{G}(p), \quad \text{for } p \in (0, 1)$$

State space:

$$E = \mathbb{N} = \{1, 2, 3, \dots\}$$

Pmf:

$$\mathbf{P}(X = k) = p(1 - p)^{k-1}, \quad k \geq 1$$

(Handwritten note: $= p q^{k-1}$)

Expected value and variance:

$$\mathbf{E}[X] = \frac{1}{p}, \quad \mathbf{Var}(X) = \frac{1 - p}{p^2}$$

Geometric random variable (2)

Use:

- Independent trials, with $\mathbf{P}(\text{success}) = p$
- $X = \#$ trials until first success $\Rightarrow X \sim \mathcal{G}(p)$

Example: dice rolling

- Set $X =$ 1st roll for which outcome = 6
- We have $X \sim \mathcal{G}(1/6) \Rightarrow E[X] = 6$

Computing some probabilities for the example: $P(k) = (1-p)^{k-1} p$

$$\mathbf{P}(X = 5) = \left(\frac{5}{6}\right)^4 \frac{1}{6} \simeq 0.08$$

$$\mathbf{P}(X \geq 7) = \left(\frac{5}{6}\right)^6 \simeq 0.33$$

$E[X]$ for $X \sim q(p)$

$$E[X] = \sum_{k=1}^{\infty} k p(k) = \sum_{k=1}^{\infty} k q^{k-1} p$$

trick

$$= \sum_{k=1}^{\infty} (k-1+1) q^{k-1} p = \sum_{k=1}^{\infty} p(k) = 1$$

$$= \sum_{k=1}^{\infty} (k-1) q^{k-1} p + \sum_{k=1}^{\infty} q^{k-1} p$$

$$= \sum_{k=1}^{\infty} (k-1) q^{k-2} q p + 1$$

$$= q \sum_{k=2}^{\infty} (k-1) q^{k-2} p + 1$$

cv: $k-1=j$

$$= q \sum_{j=1}^{\infty} j q^{j-1} p + 1$$

$$= q E[X] + 1$$

Conclusion $E[X]$ solves

$$E[X] = q E[X] + 1$$

$$\Rightarrow E[X] = \frac{1}{1-q} = \frac{1}{p}$$

Geometric random variable (3)

Computation of $\mathbf{E}[X]$: Set $q = 1 - p$. Then

$$\begin{aligned}\mathbf{E}[X] &= \sum_{i=1}^{\infty} i q^{i-1} p \\ &= \sum_{i=1}^{\infty} (i-1) q^{i-1} p + \sum_{i=1}^{\infty} q^{i-1} p \\ &= q \mathbf{E}[X] + 1\end{aligned}$$

Conclusion:

$$\mathbf{E}[X] = \frac{1}{p}$$