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# 2. Discrete random variables

$$\mathbb{N}_0 = \{0, 1, 2, 3, \ldots\}$$
  
 $\mathbb{N} = \{1, 2, 3, \ldots\}$ 

# 2.1 Probability distribution

#### Def 1: discrete random variable

$$X:\Omega\to\{x_1,x_2,x_3,\ldots\}$$

is a number resulting from a random experiment finitely or countably many possible values

$$x_1 < x_2 < x_3 < \dots$$

#### Def 2: random count

$$X:\Omega\to\mathbb{N}_0$$

is a counting result in a random experiment

| X = 0 |
|-------|
| X = 1 |
| X=2   |
| X = 3 |

Partition of  $\Omega$  caused by a r.v. X

### Def 3: probability distribution

The probability distribution of a r.v. X

is the set of probabilities for all possible values of XProbability mass function (pmf)

$$p_k = P(X = x_k), k \in \mathbb{N}, p_1 + p_2 + p_3 + \ldots = 1$$

Pmf for a random count

$$p_k = P(X = k), k \in \mathbb{N}_0, p_0 + p_1 + p_2 + \ldots = 1$$

### Ex 1: coin-die experiment

first step: a fair coin is tossed:  $P(H) = \frac{1}{2}$ ,  $P(T) = \frac{1}{2}$  second step: a die is rolled once if H or twice if T

Discrete r.v.  $D = \{ \text{total die score} \}$ 

$$p_0 = 0, p_1 = 6/72, p_2 = 7/72, \dots, p_6 = 11/72$$
  
 $p_7 = 6/72, p_8 = 5/72, \dots, p_{12} = 1/72$ 

| X = 1 | 2 | 3 | 4 | 5  | 6  | 7  |
|-------|---|---|---|----|----|----|
| X = 2 | 3 | 4 | 5 | 6  | 7  | 8  |
| X = 3 | 4 | 5 | 6 | 7  | 8  | 9  |
| X=4   | 5 | 6 | 7 | 8  | 9  | 10 |
| X = 5 | 6 | 7 |   | 9  |    |    |
| X = 6 | 7 | 8 | 9 | 10 | 11 | 12 |

#### Def 4: cumulative distribution function

$$F(k) = P(X \le k) = p_0 + p_1 + \ldots + p_k$$
  
increases from  $p_0$  to 1

Properties of cdf

$$P(X > k) = 1 - F(k) = P(X \ge k + 1)$$

$$P(k_1 < X \le k_2) = F(k_2) - F(k_1)$$

$$p_k = F(k) - F(k - 1)$$

# 2.2 Mean value and standard deviation Def 5: mean, variance, st. deviation

Mean value  $\mu$  of X or expectation E(X) is the probability mass center of X

$$\mu = \sum_{k=0}^{\infty} x_k p_k$$

$$\mu = p_1 + 2p_2 + 3p_3 + \dots \text{ for a random count}$$

Variance

$$\sigma^2 = \operatorname{Var}(X) = \operatorname{E}((X - \mu)^2)$$

is the mean squared deviation of X from its mean Standard deviation

$$\sigma = \sqrt{\operatorname{Var}(X)}$$

measures the distribution spread in same units as X

Calculate the variance by 
$$\sigma^2 = E(X^2) - \mu^2$$

# Properties of Expectation and Variance

$$E(X + Y) = E(X) + E(Y)$$

$$E(c \cdot X) = c \cdot E(X), \operatorname{Var}(c \cdot X) = c^2 \cdot \operatorname{Var}(X)$$

$$Eg(X) = \sum_{k=0}^{\infty} g(x_k) p_k, E(X^i) = \sum_{k=0}^{\infty} (1 - F(k)) \text{ for a random count}$$

If 
$$X$$
 and  $Y$  are independent, then  $E(XY)=E(X)E(Y)$ ,  $Var(X+Y)=Var(X)+Var(Y)$ 

# Ex 2: students' grades

Compare three grade distributions:

| araac                  | 2  |    |    |    | Total |
|------------------------|----|----|----|----|-------|
| Student A              | 25 | 25 | 25 | 25 | 100%  |
| Student A<br>Student B | 40 | 10 | 10 | 40 | 100%  |
| Student C              | 10 | 40 | 40 | 10 | 100%  |

| X                 | $\mathrm{E}(X)$ | $\mathrm{E}(X^2)$ | $\operatorname{Var}(X)$ | $\sigma_X$ |
|-------------------|-----------------|-------------------|-------------------------|------------|
| Student A's grade | 3.5             | 13.5              | 1.25                    | 1.12       |
| Student B's grade | 3.5             | 14.1              | 1.85                    | 1.36       |
| Student C's grade | 3.5             | 12.9              | 0.65                    | 0.81       |

#### 2.3 Discrete uniform distribution

Discrete uniform distr. with parameter N

$$X \sim U(N), N \in \mathbb{N}$$
  
 $p_k = \frac{1}{N}, k = 1, ..., N$   
 $\mu = \frac{N+1}{2}, \sigma^2 = \frac{N^2-1}{12}$ 

### Ex 3: systematic search

Open a lock by trying codes: 0000, 0001, 0002, ... number of trials required:  $X \sim \text{U}(10000)$   $\mu = 5000.5 \text{ trials}$   $\sigma^2 = 8.3 \cdot 10^6 \text{ squared trials}$  $\sigma = 2886.8 \text{ trials}$ 

#### 2.4 Binomial distribution

Binomial distribution with parameters n and p

$$X \sim \text{Bin}(n, p), n \in \mathbb{N}, 0 
$$p_k = \binom{n}{k} p^k q^{n-k}, k = 0, 1, \dots, n, q = 1 - p$$

$$\mu = np$$

$$\sigma^2 = npq, \sigma = \sqrt{npq}$$$$

#### Def 6: Bernoulli trials

independently repeated experiment with two possible outcomes: success or failure Number of successes in n Bernoulli trials  $X \sim \text{Bin}(n,p)$  p is the probability of success q=1-p is the probability of failure

If 
$$X \sim \text{Bin}(n, p)$$
, then  $X = I_1 + \ldots + I_n$   
where  $I_1, \ldots, I_n$  are independent with  $P(I_j = 1) = p$   
 $P(I_j = 0) = q$ ,  $E(I_j) = p$ ,  $Var(I_j) = pq$ 

# Ex 4: sampling with replacement

Consider a box with white and black balls:

N=30 the total number of balls  $p=\frac{1}{3}$  the proportion of black balls in the box Randomly sample n=5 balls with replacement number of black balls in the sample  $X \sim \text{Bin}(5,\frac{1}{3})$   $P(\text{BBBWW}) = p^3q^2 = 0.0165$   $P(X=3) = \binom{5}{3} \cdot p^3q^2 = 0.165$ 

#### Ex 5: Ehrenfest model of diffusion

Suppose n molecules of a gas are in a container divided into two equal parts by a permeable membrane  $X_t = \text{number of molecules}$  in the left part at time t Transition probabilities

$$P(X_{t+1} = k - 1 | X_t = k) = k/n$$

$$P(X_{t+1} = k + 1 | X_t = k) = (n - k)/n$$
Equilibrium distribution  $p_k = P(X_t = k)$ 

$$p_k = p_{k-1}(n - k + 1)/n + p_{k+1}(k + 1)/n$$

$$p_k = \binom{n}{k} 2^{-n}, k = 0, 1, ..., n$$

Equilibrium distribution is Bin(n, 1/2) each molecule chooses one of two parts independently at random

# 2.5 Hypergeometric distribution

Hypergeometric distribution with parameters N, n, p

$$X \sim \operatorname{Hg}(N, n, p)$$

$$n, N, (Np) \in \mathbb{N}, n \le N, 0$$

$$p_k = \frac{\binom{Np}{k}\binom{Nq}{n-k}}{\binom{N}{n}}, \max(n-Nq,0) \le k \le \min(n,Np)$$

$$\mu = np$$

$$\sigma^2 = npq(1 - \frac{n-1}{N-1})$$

Sampling without replacement

N = the total number of balls in the box

p = initial proportion of black balls in the box

X = number of black balls in the sample of size n

$$X = I_1 + \ldots + I_n$$
 with  $P(I_j = 1) = p$ ,  $P(I_j = 0) = q$ 

Reduced variance due to

negative dependence between  $I_1, \ldots, I_n$ 

the more black balls are drawn

the less chances to see another black ball

The finite population correction  $(1 - \frac{n-1}{N-1})$  is negligible when the sample fraction  $\frac{n}{N}$  is small

# Ex 6: sampling without replacement

5 balls sampled without replacement

from a box with 10 black and 20 white balls

 $\binom{30}{5}$  unordered samples are equally likely

Division rule:

$$P(3 \text{ black} + 2 \text{ white}) = \frac{\binom{10}{3}\binom{20}{2}}{\binom{30}{5}} = \frac{120 \cdot 190}{142506} = 0.16$$

### Ex 7: aspirin teatment

placebo group: 11034 individuals, 189 heart attacks aspirin group: 11037 individuals, 104 heart attacks Statistical model

X = number of heart attacks in the placebo groupwithout aspirin effect  $X \sim \text{Hg}(N, n, p)$ 

$$N = 22071, n = 293, p = \frac{11034}{22071} = 0.4999$$

$$N = 22071, n = 293, p = \frac{11034}{22071} = 0.4999$$

$$P(X = 189) = \frac{\binom{11034}{189}\binom{11037}{104}}{\binom{22071}{293}} = 0.00000015$$

Even the maximal probability is small

$$P(X = 146) = P(X = 147) = 0.0468$$

A different proportion

 $P(X \ge 189)$  would be more informative

#### 2.6 Geometric distribution

Geometric distribution with parameter p

$$X \sim \text{Geom}(p), 0$$

$$p_k = pq^{k-1}, k \in \mathbb{N}, \operatorname{cdf} F(k) = 1 - q^k$$

$$\mu = \frac{1}{p}, \ \sigma^2 = \frac{q}{p^2}$$

Bernoulli trials with probability of success p

X = number of trials until the first success

Skewed (non-symmetric) pmf shape

$$p_{k+1} = p_k \cdot q$$

Lack of memory property for the geometric distribution

$$P(X > t + k | X > t) = \frac{P(X > t + k)}{P(X > t)} = \frac{q^{t+k}}{q^t} = P(X > k)$$

### Ex 8: birthday problem

Number of, people asked until the same birthday as yours

$$X \sim \text{Geom}(1/365)$$

$$P(X > 253) = (\frac{364}{365})^{253} = 0.5$$

mean of X is 365, median of X is 253

#### Ex 9: random search

Try the lock codes at random

number of trials required 
$$X \sim \text{Geom}(10^{-4})$$

$$\mu = 10000 \text{ trials}, \, \sigma \approx 10000$$

$$P(X > 10000) = (0.9999)^{10000} = 0.37 \approx e^{-1}$$

# 2.7 Negative binomial distribution

Negative binomial distribution with parameters r, p

$$X \sim \text{Nb}(r, p), r \in \mathbb{N}, 0$$

$$p_k = {k-1 \choose r-1} p^r (1-p)^{k-r}, k = r, r+1, r+2, \dots$$

$$\mu = \frac{r}{p}, \ \sigma^2 = \frac{rq}{p^2}$$

Bernoulli trials with probability of success p

X = number of trials until the r-th success

$$X = Y_1 + \ldots + Y_r$$
 with independent  $Y_i \sim \text{Geom}(p)$ 

### 2.8 Poisson distribution

Poisson distribution with parameter  $\lambda$ 

$$X \sim \text{Pois}(\lambda), \lambda > 0$$

$$p_k = \frac{\lambda^k}{k!} e^{-\lambda}, k \in \mathbb{N}_0$$

computational formula:  $p_{k+1} = p_k \cdot \frac{\lambda}{k+1}$ 

$$\mu = \sigma^2 = \lambda$$

### Poisson approximation of the Binomial distr

Poisson distribution is a distribution law of rare events small p and large n (jackpot wins, accidents)

$$Bin(n,p) \approx Pois(np) \text{ if } n \geq 100, p \leq 0.01$$

Exact meaning: for any fixed 
$$k \in \mathbb{N}_0$$

$$\binom{n}{k} p^k (1-p)^{n-k} \sim \frac{n^k}{k!} p^k e^{-np} \to \frac{\lambda^k}{k!} e^{-\lambda}$$
as  $np \to \lambda$ 

# Poisson process of radioactive disintegrations

Radioactivity as a flow of Bernoulli trials

p = probability of a disintegration per a millisecond number of disintegrations during t hours

$$X_t \sim \text{Pois}(\lambda t)$$
, where  $\lambda = 1440000p$ 

Poisson process  $\{X_t\}$  counts disintegrations occurring at the rate  $\lambda$  disintegrations per hour Other examples of rates

3 asteroids per MY hit the Earth, MY = million years 5 replacements per amino acid per 1000 MY

### Ex 10: cystic fybrosis

proportion of affected people p = 1/3000

 $X = \#\{\text{affected in a random sample of size } n = 6000\}$ Poisson approximation:

P(X = 3) = 
$$\binom{6000}{3} (\frac{1}{3000})^3 (\frac{2999}{3000})^{5997} \approx \frac{2^3}{3!} e^{-2} = 0.180$$
  
P(X = 1) =  $2e^{-2} = 0.271$   
P(X \le 3) =  $e^{-2} + 2e^{-2} + \frac{2^2}{2}e^{-2} + \frac{2^3}{6}e^{-2} = 0.857$