

CS 107 2-27-24

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• midterm 2: Thurs Feb. 29

• skip 4.4

4.5 sum of a random # of r.v.s

Let X_1, X_2, X_3, \dots be an infinite seq. of independent, identically distributed r.v.s abbreviated: (iid)

This means they all have the same PDF, (or PMF) which we write

$$f_X(x) \quad \text{or} \quad P_X(x)$$

Let N be a r.v. with values in positive integers $\{1, 2, 3, \dots\}$.

Assume: $\{N, X_1, X_2, \dots\}$

is independent. Now define

$$Y = X_1 + X_2 + \dots + X_N$$

we denote by $E[X]$ and $\text{var}(X)$ the common mean & variance of all X_i .

Goal: find $E[Y]$ and $\text{var}(Y)$.

Note:

$$E[Y | N=n]$$

$$= E[X_1 + X_2 + \dots + X_N | N=n]$$

$$= E[X_1 + X_2 + \dots + X_n | N=n]$$

$$= E[X_1 + X_2 + \dots + X_n] \quad \left\{ \begin{array}{l} \text{by} \\ \text{indep.} \end{array} \right.$$

$$= n E[X]$$

Replace n by N to get

$$E[Y|N] = N \cdot E[X]$$

By law of iterated expectation

$$E[Y] = E[E[Y|N]]$$

$$= E[N \cdot E[X]]$$

$$= E[N] \cdot E[X] \leftarrow$$

Also

$$\text{Var}(Y | N=n)$$

$$= \text{Var}(X_1 + X_2 + \dots + X_N | N=n)$$

$$= \text{Var}(X_1 + \dots + X_n | N=n)$$

$$= \text{Var}(X_1 + \dots + X_n) \quad \text{by indep.}$$

$$= \text{Var}(X_1) + \dots + \text{Var}(X_n) \quad \text{by indep.}$$

$$= n \text{Var}(X)$$

$$\therefore \text{Var}(Y | N) = N \cdot \text{Var}(X)$$

By the law of total variance

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$$\text{Var}(Y) = E[\text{Var}(Y|N)] + \text{Var}(E[Y|N])$$

$$= E[N \text{Var}(X)] + \text{Var}(N E[X])$$

$$= E[N] \cdot \text{Var}(X) + E[X]^2 \cdot \text{Var}(N)$$

EX. Suppose have 2 coins

$$\text{coin 1: } P(\text{head}) = p$$

$$\text{coin 2: } P(\text{head}) = q$$

□

Flip coin 1 until 1st head, let N be # flips. note N is geometric with Param. p .

Flip coin 2 N times and count # heads, let Y be # heads in 2nd sequence.

observe

$$Y = X_1 + X_2 + \dots + X_N$$

where each X_i is Bernoulli with parameter q . also the X_i are indep. of each other and of N .

let $E[X]$, $\text{Var}(X)$ denote
their common mean $\frac{1}{p}$, Var .

we have

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

and

$$E[X] = q, \quad \text{Var}(X) = q(1-q).$$

Thus

$$E[Y] = E[N] \cdot E[X] = \frac{q}{p}$$

and

$$\text{Var}(Y) = E[N] \text{Var}(X) + E[X]^2 \text{Var}(N)$$

$$= \frac{q(1-q)}{p} + \frac{q^2(1-p)}{p^2}$$

6.1 The Bernoulli Process

Defn

The Bernoulli Process is a sequence

$$X_1, X_2, \dots$$

of indep. Bernoulli r.v.s with

$$P(X_t = 1) = p \quad (\text{heads, success, arrival})$$

$$P(X_t = 0) = 1 - p \quad (\text{tails, failure, non-arrival})$$

we can think of this as

- one experiment performed ∞ -many times: $\Omega = \{0, 1\}$

or

- a single experiment with $\Omega = \{ \text{infinite bit strings} \}$

Have 2 basic questions:

- ① given # of trials n , how many successes!

answer: Binomial (n, p)

② how many trials until 1st success ?

answer : Geometric (p)

Memorylessness

Let T be time # trials to 1st success. Then T is

Geometric(p), i.e.

$$P_T(t) = (1-p)^{t-1} p$$

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

we observe

$$X_1, X_2, \dots, X_n$$

are all failures. Then the event $\{T > n\}$ has occurred.

what is distribution of $T-n$,
the time left until 1st head.

$$P(T-n = t | T > n) = \frac{P(T-n = t, T > n)}{P(T > n)}$$

$$= \frac{P(T = n+t, T > n)}{P(T > n)}$$

$$= \frac{P(T = n+t)}{P(T > n)} \quad \left\{ \begin{array}{l} \text{if } A \subseteq B \\ \text{then } A \cap B = A \end{array} \right.$$

$$= \frac{(1-p)^{n+t-1} \cdot p}{(1-p)^n}$$

$$= \frac{\cancel{(1-p)^n} \cdot (1-p)^{t-1} \cdot p}{\cancel{(1-p)^n}}$$

$$= (1-p)^{t-1} \cdot p$$

$$= P(T = t) = p_{II}(t)$$

as expected, independence implies memoryless.

Inter-arrival times

let

\overline{T}_1 = time to 1st arrival

\overline{T}_2 = from 1st arrival to 2nd

⋮

\overline{T}_k = " (k-1)th " " kth

then

$$Y_k = \overline{T}_1 + \overline{T}_2 + \dots + \overline{T}_k$$

= time to kth arrival