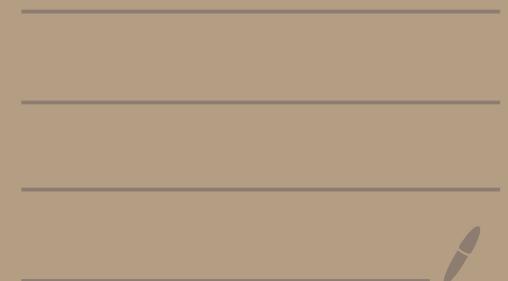


PTS - Chapter 3



3 Special Random Variables

3.1 Bernoulli and Binomial

A Bernoulli experiment has only 2 outcomes

$$X = \begin{cases} 1 & \text{if success} \\ 0 & \text{if failure} \end{cases}$$

Prob of X :

$$P[X = 1] = p, \quad \text{for some } 0 \leq p \leq 1$$

$$P[X = 0] = (1-p)$$

Let X be a Bernoulli variable with probability p . Then

$$E[X] = 1 \cdot p + 0 \cdot (1-p) = p$$

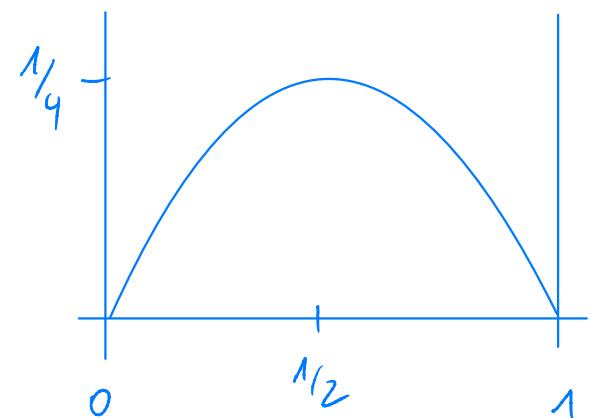
Note: $X^2 = X$

$$E[X^2] = E[X] = p$$

$$\text{Var}(X) = E[X^2] - E[X]^2$$

$$= p - p^2 = p(1-p)$$

maximal for $p = \frac{1}{2}$



Graph of $p(1-p)$,
maximal for $p = \frac{1}{2}$

Let us repeat a Bernoulli experiment with RV X .

Suppose X_1, \dots, X_n, \dots are i.i.d. $\text{Bern}(p)$ RVs.

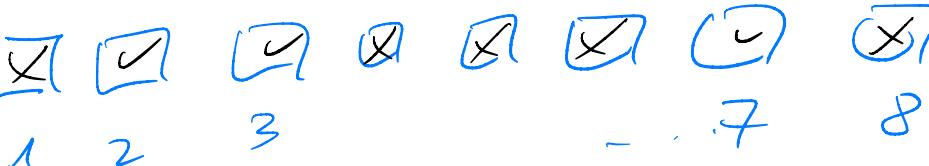
Then

$y_n := \sum_{i=1}^n X_i$ counts the number of successes

Possible values of $y_n = 0, 1, \dots, n$

$$P[Y_n = i] = \binom{n}{i} p^i (1-p)^{n-i}$$

Boxes to record
success


1 2 3 . . 7 8 3 successes

- # ways to get 3 successes? $\binom{8}{3}$

- Probability of such an outcome: $p^3 (1-p)^{8-3}$

Let us repeat a Bernoulli experiment with RV X n times

Suppose X_1, \dots, X_n are i.i.d. $\text{Bern}(p)$ RVs.

Then

$Y_n := \sum_{i=1}^n X_i$ counts the number of successes

The possible values of Y_n are $0, 1, 2, \dots, n$.

$$P[Y_n = i] = \binom{n}{i} p^i (1-p)^{n-i}$$

1 2 3 . . . 7 8 3 successes

Boxes to record
success

• #ways to get 3 successes? $\binom{8}{3}$

• Probability of such an outcome: $p^3(1-p)^{8-3}$

Note :

$$(p+q)^n = \sum_{i=0}^n \binom{n}{i} p^i q^{n-i}$$

according to the **binomial theorem**.

Suppose $q = 1-p$. Then

$$1 = (p + (1-p))^n = \sum_{i=0}^n \binom{n}{i} p^i (1-p)^{n-i}.$$

This shows that

$$p_i := \binom{n}{i} p^i (1-p)^{n-i}, \quad 0 \leq i \leq n,$$

is a **probability mass function**.

we say that y_n is distributed according to the binomial distribution with parameters n and p , written

$$y_n \sim B(n, p).$$

we calculate mean and variance:

$$E[y_n] = E\left[\sum_{i=1}^n x_i\right] = \sum_{i=1}^n E[x_i] = \sum_{i=1}^n p = np$$

$$\begin{aligned} \text{Var}(y_n) &= \text{Var}\left(\sum_{i=1}^n x_i\right) = \sum_{i=1}^n \text{Var}(x_i) \\ &= \sum_{i=1}^n p(1-p) = np(1-p) \end{aligned}$$

Example: A satellite system has 4 components and functions if at least 2 are working.

Each component is independently working with probability $P = 0.6$.

What is the probability that the system functions?

$$P[\text{System functions}]$$

$$= 1 - P[\text{System doesn't function}]$$

$$= 1 - p_0$$

$$p_0 = P[\text{all components fail}] + P[\text{exactly 3 components fail}]$$

$$= P[\text{no component functions}] + P[\text{exactly 1 comp. functions}]$$

$$= \binom{4}{0} p^0 (1-p)^4 + \binom{4}{1} p^1 (1-p)^3$$

$$= 1 \cdot 0.4^4 + 4 \cdot 0.6 \cdot 0.4^3 = 0.1792$$

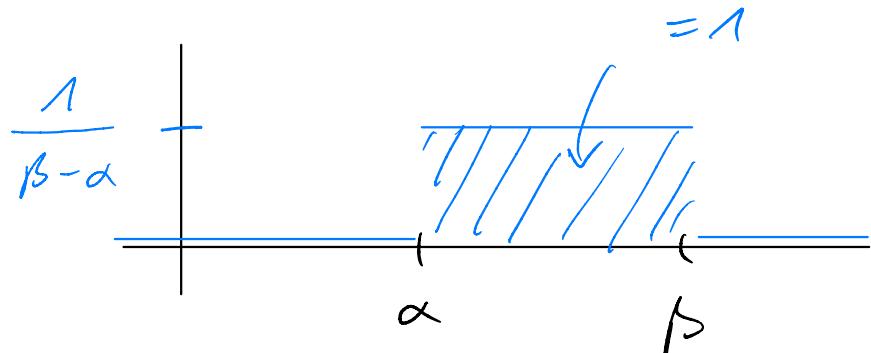
3.3 Uniform Random Variables

A continuous RV X is uniformly distributed if there is an interval $[\alpha, \beta]$ so that

- X takes only values in $[\alpha, \beta]$
- all values are equally probable.

This means, X has the density f where

$$f(x) = \begin{cases} \frac{1}{\beta - \alpha}, & \alpha \leq x \leq \beta \\ 0 & \text{otherwise.} \end{cases}$$



We write $X \sim U[\alpha, \beta]$.

What about mean and variance?

We first determine mean and variance for the simple case that $X \sim U[0,1]$. Then

$$E[X] = \int_0^1 x \cdot 1 dx = \left[\frac{x^2}{2} \right]_0^1 = \frac{1}{2}$$

$$E[X^2] = \int_0^1 x^2 \cdot 1 dx = \left[\frac{x^3}{3} \right]_0^1 = \frac{1}{3}$$

Hence,

$$\begin{aligned} \text{Var}(X) &= E[X^2] - E[X]^2 \\ &= \frac{1}{3} - \left(\frac{1}{2}\right)^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12} \end{aligned}$$

Suppose now that $y \sim U[\alpha, \beta]$.

Then $\xi := \frac{y - \alpha}{\beta - \alpha}$ is $U[0, 1]$ -distributed and

$$y = (\beta - \alpha) \xi + \alpha.$$

Therefore,

$$\begin{aligned} E[y] &= (\beta - \alpha) E[\xi] + \alpha = \frac{\beta - \alpha}{2} + \alpha \\ &= \frac{\alpha + \beta}{2} \end{aligned}$$

$$\text{Var}(y) = (\beta - \alpha)^2 \text{Var}(\xi) = \frac{(\beta - \alpha)^2}{12}$$

Mean of $\text{U}[\alpha, \beta]$: Brute-Force Calculation

This is a calculation of mean and variance following directly the definition. Compare this to our approach of (i) solving a simple variant of the problem and (ii) reducing complex cases to the simple one.

Also:

$$\begin{aligned} E[X] &= \int_{\alpha}^{\beta} x \cdot \frac{1}{\beta-\alpha} dx = \frac{1}{\beta-\alpha} \left[\frac{x^2}{2} \right]_{\alpha}^{\beta} \\ &= \frac{1}{\beta-\alpha} \frac{1}{2} (\beta^2 - \alpha^2) = \frac{(\beta+\alpha)(\beta-\alpha)}{2(\beta-\alpha)} \\ &= \frac{\alpha + \beta}{2} \end{aligned}$$

Variance of $E[\alpha, \beta]$: Brut-Force Calculation

$$\begin{aligned} E[X^2] &= \int_{\alpha}^{\beta} x^2 \frac{1}{\beta-\alpha} dx = \frac{1}{\beta-\alpha} \left[\frac{x^3}{3} \right]_{\alpha}^{\beta} \\ &= \frac{1}{3} \frac{\beta^3 - \alpha^3}{\beta - \alpha} = \frac{1}{3} (\beta^2 + \alpha\beta + \alpha^2) \end{aligned}$$

$$\begin{aligned} \text{Var}(X) &= E[X^2] - E[X]^2 \\ &= \frac{\beta^2 + \alpha\beta + \alpha^2}{3} - \frac{(\alpha + \beta)^2}{2^2} \\ &= \frac{4\beta^2 + 4\alpha\beta + 4\alpha^2 - 3\alpha^2 - 6\alpha\beta - 3\beta^2}{12} \\ &= \frac{1}{12} (\beta^2 - 2\alpha\beta + \alpha^2) = \frac{(\beta - \alpha)^2}{12} \end{aligned}$$