Special continuous random variables

January 14, 2021





Exponential RV

X is an exponential RV with parameter λ if its pdf is

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \ge 0\\ 0, & otherwise \end{cases}$$

Denote $X \sim \mathcal{E}(\lambda)$





Use exponential RV to

model arrival time of something, e.g. the amount of time until a piece of equipment breaks down, until a light bulb burns out, or until an accident occurs ...





Mean and Variance of $\mathcal{E}(\lambda)$

 $X \sim \mathcal{E}(\lambda)$ then

- $E(X) = \frac{1}{\lambda}$
- $\operatorname{Var}(X) = \frac{1}{\lambda^2}$



Example

The time until a small meteorite first lands anywhere in the Sahara desert is modeled as an exponential random variable with a mean of 10 days. The time is currently midnight. What is the probability that a meteorite first lands some time between 6am and 6pm of the first day?





• $X \sim \lambda$: time until a first lands

•
$$E(X) = 10 \Rightarrow \frac{1}{\lambda} = 10 \text{ or } \lambda = \frac{1}{10}$$

• pdf of X

$$f(x) = \begin{cases} \frac{1}{10}e^{-\frac{1}{10}x}, & x \ge 0\\ 0, & otherwise \end{cases}$$



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- $X \sim \lambda$: time until a first lands
- $E(X) = 10 \Rightarrow \frac{1}{\lambda} = 10 \text{ or } \lambda = \frac{1}{10}$
- pdf of *X*

$$f(x) = \begin{cases} \frac{1}{10}e^{-\frac{1}{10}x}, & x \ge 0\\ 0, & otherwise \end{cases}$$





• between 6am to 6pm of the first day = $\{\frac{1}{4} \le X \le \frac{3}{4}\}$ (day)

$$P(\frac{1}{4} \le X \le \frac{3}{4}) = \int_{1/4}^{3/4} \frac{1}{10} e^{-\frac{1}{10}x} dx$$

$$\approx .0476$$

Normal RV

Continuous RV X is said to be normally distributed or Gaussian with parameter μ and σ^2 if its pdf is

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-(x-\mu)^2/2\sigma^2}$$

for
$$-\infty < x < \infty$$

Denote $X \sim \mathcal{N}(\mu, \sigma^2)$.



Mean and variance of $\mathcal{N}(\mu, \sigma^2)$

$$X \sim \mathcal{N}(\mu, \sigma^2)$$

- $\bullet \ E(X) = \mu$
- $Var(X) = \sigma^2$



Bell shape, symmetric about the mean

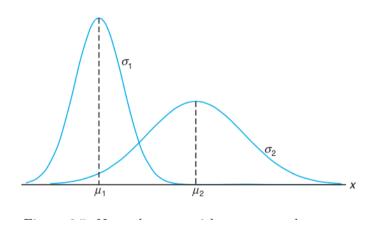


Figure: $\mu_1 < \mu_2$, $\sigma_1 < \sigma_2$



Application

- Normal distibution is the most widely used distribution
- Many random phenomena obey a normal distribution
- Ex: the height and weight of a person, accuracy of shots from a gun... Link





Good approximation

- to approximate Binomial (n, p) when n is large
- *limiting distribution* of sample mean ...**broad base** for statistic inference (estimation and hypothesis testing), analysis of variance





Standard normal distribution

- $Z \sim \mathcal{N}(0, 1)$ is standard normal distribution
- pdf

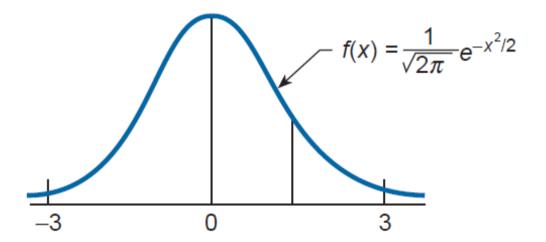
$$f(x) = \frac{1}{\sqrt{2\pi}}e^{-\frac{x^2}{2}}$$

• cdf

$$\Phi(x) = P(Z \le x) = \int_{-\infty}^{x} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$











Compute probability of standard normal distribution

- Calculator
- Look up values in Normal Probability Table





Standard ormal probability table

735

Table A.3 Normal Probability Table

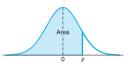


Table A.3 Areas under the Normal Curve

z	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-3.4	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0002
-3.3	0.0005	0.0005	0.0005	0.0004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0003
-3.2	0.0007	0.0007	0.0006	0.0006	0.0006	0.0006	0.0006	0.0005	0.0005	0.0005
-3.1	0.0010	0.0009	0.0009	0.0009	0.0008	0.0008	0.0008	0.0008	0.0007	0.0007
-3.0	0.0013	0.0013	0.0013	0.0012	0.0012	0.0011	0.0011	0.0011	0.0010	0.0010
-2.9	0.0019	0.0018	0.0018	0.0017	0.0016	0.0016	0.0015	0.0015	0.0014	0.0014
-2.8	0.0026	0.0025	0.0024	0.0023	0.0023	0.0022	0.0021	0.0021	0.0020	0.0019
-2.7	0.0035	0.0034	0.0033	0.0032	0.0031	0.0030	0.0029	0.0028	0.0027	0.0026
-2.6	0.0047	0.0045	0.0044	0.0043	0.0041	0.0040	0.0039	0.0038	0.0037	0.0036
-2.5	0.0062	0.0060	0.0059	0.0057	0.0055	0.0054	0.0052	0.0051	0.0049	0.0048





Example

 $P(Z \le -2.54)$



Solution 1 - Calculator

$$P(Z \le -2.54) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{2.54} e^{-\frac{x^2}{2}} dx$$
$$= \lim_{a \to -\infty} \frac{1}{\sqrt{2\pi}} \int_{a}^{-2.54} e^{-\frac{x^2}{2}} dx$$

Substitute a by -10, -30, -50 ... and find the limit





Solution 2 - Look up the table value of normal probability

- \bigcirc look up -2 in the first column
- 2 look up 54 in the first row
- 3 Intersection of the corresponding row and column

$$P(Z < -2.54) = .0055$$





Practice

Find

- P(Z > 2.33)
- **2** P(-1.65 < Z < 1.65)
- **3** z such that P(Z > z) = .95

$(1 - \alpha)$ percentile

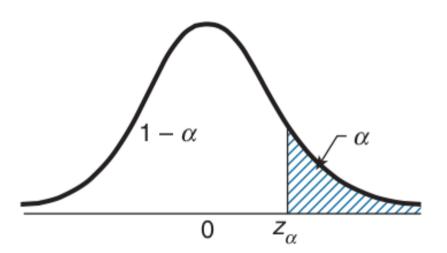
 z_{α} is the $100(1-\alpha)$ percentile if

$$P(Z \le z_{\alpha}) = 1 - \alpha$$

or

$$P(Z > z_{\alpha}) = \alpha$$

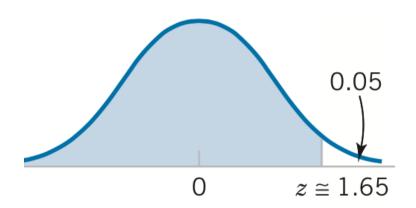








Example

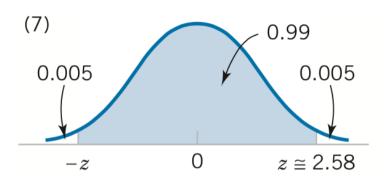


$$z_{.05} = 1.65$$





Example



$$z_{.005} = 2.58$$

$$z_{.995} = -z_{.005} = -2.58$$





Properties of percentile

$$z_{1-\alpha} = -z_{\alpha}$$

because of the symmetry about y- axis of standard normal distribution





Normality is Preserved by Linear Transformations

If
$$X \sim \mathcal{N}(\mu, \sigma^2)$$
 then $Y = aX + b \sim \mathcal{N}(a\mu + b, a^2\sigma^2)$



Standardize a normal distribution

If
$$X \sim \mathcal{N}(\mu, \sigma^2)$$
 then

$$Z = \frac{X - \mu}{\sigma} \sim \mathcal{N}(0, 1)$$



cdf calculation for normal

If $X \sim \mathcal{N}(\mu, \sigma^2)$ then

$$P(X \le x) = P(\frac{X - \mu}{\sigma} \le \frac{x - \mu}{\sigma})$$
$$= P(Z \le \frac{x - \mu}{\sigma})$$

and

$$P(a \le X \le b) = P(Z < \frac{b-\mu}{\sigma}) - P(Z < \frac{a-\mu}{\sigma})$$





Example

The annual snowfall at a particular geographic location is modeled as a normal random variable with a mean of $\mu=60$ inches, and a standard deviation of $\sigma=20$. What is the probability that this year's snowfall will be at least 80 inches?





• Snowfall $X \sim \mathcal{N}(\mu, \sigma^2)$ with $\mu = 60, \sigma = 20$

$$P(X \ge 80) = 1 - P(X < 80)$$

$$= 1 - P(Z < \frac{80 - \mu}{\sigma})$$

$$= 1 - P(Z < \frac{80 - 60}{20})$$

$$= 1 - P(Z < 1) = 1 - .8413 = .1687$$



Binomial and normal

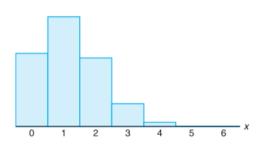


Figure 6.24: Histogram for b(x; 6, 0.2).

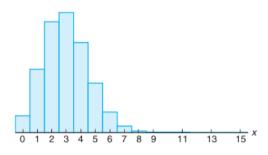


Figure 6.25: Histogram for b(x; 15, 0.2).



Binomial approximation

- Suppose $Y \sim \text{Bino}(n, p)$ where n is large and np is not too small
- *Y* can be approximated by $X \sim \mathcal{N}(np, np(1-p))$
- *Y* is discrete, *X* is continuous
- so we have to "fill the gap"





"Fill the gap" - midpoint rule

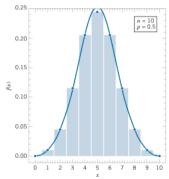


Figure 4-19 Normal approximation to the binomial distribution.

$$P(Y = i) \approx P(i - \frac{1}{2} < X < i + \frac{1}{2})$$





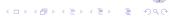
Continuity correction

To approximate a binomial probability of $X \hookrightarrow Bin(n,p)$ with a normal distribution, a **continuity correction** is applied as follows:

$$P(X \le x) = P(X \le x + 0.5) \approx P\left(Z \le \frac{x + 0.5 - np}{\sqrt{np(1 - p)}}\right)$$

The approximation is good for np > 5 and n(1-p) > 5.





Example

Toss a fair coin 40 times. Y is the number of heads. Calculate P(Y = 20) using normal approximation and direct computation.



Approximate *Y* by $X \sim \mathcal{N}(20, 10)$

$$P(Y = 20) \approx P(19.5 < X < 20.5)$$

$$= P(\frac{19.5 - 20}{\sqrt{10}} < Z < \frac{20.5 - 20}{\sqrt{10}})$$

$$= P(Z < .16) - P(Z < -.16)$$

$$= .1272$$

Exact value

$$P(Y = 20) = {40 \choose 20} (.5)^{40} = .1254$$



Example

The ideal size of a first-year class at a particular college is 150 students. The college, knowing from past experience that, on the average, only 30 percent of those accepted for admission will actually attend, uses a policy of approving the applications of 450 students. Compute the probability that more than 150 first-year students attend this college.





X: number of attending students $X \sim \text{Bino}(450, .3) \approx \mathcal{N}(135, 94.5)$

$$P(X > 150) = P(X \ge 150.5)$$

$$\approx P(Z \ge \frac{150.5 - 135}{\sqrt{94.5}})$$

$$= 1 - P(Z < 1.59)$$

$$= .0559 \approx 5.6\%$$