# Week 06 notes

# Continuous families: summary

## 01 Theory

**△** Memorize this info!

#### **Uniform:** $X \sim \text{Unif}([a, b])$

- All times  $a \leq t \leq b$  equally likely.
- $f_X(t) = \frac{1}{b-a}$
- $E[X] = \frac{a+b}{2}$
- $Var[X] = \frac{1}{12}(b-a)^2$

#### **Exponential:** $X \sim \operatorname{Exp}(\lambda)$

- Measures wait time until first arrival.
- $f_X(t) = \lambda e^{-\lambda t}$
- $E[X] = \frac{1}{\lambda}$
- $\operatorname{Var}[X] = \frac{1}{\lambda^2}$

**Erlang:**  $X \sim \operatorname{Erlang}(\ell, \lambda)$ 

- Measures wait time until  $\ell^{\rm th}$  arrival.
- $f_X(t)=rac{\lambda^\ell}{(\ell-1)!}t^{\ell-1}e^{-\lambda t}$
- $E[X] = \frac{\ell}{\lambda}$
- $\operatorname{Var}[X] = \frac{\ell}{\lambda^2}$

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

- Limiting distribution of large sums.
- $f_X(x)=rac{1}{\sqrt{2\pi\sigma^2}}e^{-(x-\mu)^2/2\sigma^2}$
- $E[X] = \mu$
- $\operatorname{Var}[X] = \sigma^2$

## Normal distribution

### **02 Theory**

A variable X has a **normal distribution**, written  $X \sim \mathcal{N}(\mu, \sigma^2)$ , when it has PDF given by:

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

The standard normal is  $Z \sim \mathcal{N}(0, 1)$  and its PDF is usually denoted by  $\varphi$ :

$$arphi(x)=rac{1}{\sqrt{2\pi}}e^{-x^2/2}$$

The standard normal CDF is **denoted by**  $\Phi(z)$ :

$$\Phi(z) = \int_{-\infty}^z rac{1}{\sqrt{2\pi}} e^{-u^2/2}\, dx$$

• To show that  $\varphi(x)$  is a valid probability density, we must show that  $\int_{-\infty}^{+\infty} \varphi(x) \, dx = 1$ .

- This calculation is *not trivial*; it requires a double integral in polar coordinates!
- There is *no explicit antiderivative* of  $\varphi$ 
  - A computer is needed for numerical calculations.
  - A chart of approximate values of  $\Phi$  is provided for exams.
- Check that E[Z] = 0
  - Observe that  $x\varphi(x)$  is an *odd function*, i.e. symmetric about the y-axis.
  - One must also verify that the integral converges.
- Check that  $\operatorname{Var}[Z] = 1$ 
  - Since  $\mu = E[Z] = 0$ , we find:

$${
m Var}[Z] = E[Z^2] \quad \gg \gg \;\; rac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} x^2 e^{-x^2/2} \, dx =: R$$

• Use integration by parts to compute that I = 1.

Generalized normal distributions are related to standard normal distributions by linear transformations. The generalized normal  $X \sim \mathcal{N}(\mu, \sigma^2)$  satisfies  $X \sim \sigma Z + \mu$ , where  $Z \sim \mathcal{N}(0, 1)$ .

Let us check directly that  $X \sim \sigma Z + \mu$  by showing their PDFs are equal. Computing the CDF of *X*, we find:

$$egin{aligned} F_X(x) &= P[X \leq x] \ &= P[\sigma Z + \mu \leq x] \ &= P[Z \leq rac{x-\mu}{\sigma}] \ &= \Phi\left(rac{x-\mu}{\sigma}
ight) \end{aligned}$$

Then we can find the PDF of X by differentiating  $F_X$ :

$$egin{aligned} f_X(x) &= rac{d}{dx} F_X(x) \ &= rac{d}{dx} \Phi\left(rac{x-\mu}{\sigma}
ight) \ &= rac{1}{\sigma} arphi\left(rac{x-\mu}{\sigma}
ight) \ &= rac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/2\sigma^2} \end{aligned}$$

Now since we know  $X \sim \sigma Z + \mu$ , we can infer that  $E[X] = \mu$  and  $\operatorname{Var}[X] = \sigma^2$ .

## **03 Illustration**

 $\equiv$  Example - Basic generalized normal calculation

Suppose  $X \sim \mathcal{N}(-3, 4)$ . Find  $P[X \ge -1.7]$ .

#### Solution

First write X as a linear transformation of Z:

 $X\sim 2Z-3$ 

Then:

$$X \geq -1.7 \quad \Longleftrightarrow \quad Z \geq 0.65$$

#### $\equiv$ Example - Gaussian: interval of 2/3

Find the number a such that  $P[Z \in [-a, a]] = 2/3$ .

#### Solution

First convert the question:

$$egin{array}{lll} Pig[Z\in [-a,a]ig] &\gg &\gg &F_Z(a)-F_Z(-a) \ &\gg &\gg &\Phi(a)-\Phi(-a)b \ &\gg &\gg &2\Phi(a)-1 \ &\gg &\gg &rac{2}{2} \end{array}$$

Solve for  $\Phi(a) = \frac{5}{6}$ . Use a  $\Phi$  table to conclude  $a \approx 0.97$ .

#### $\equiv$ Example - Heights of American males

Suppose that the height of an American male in inches follows the normal distribution  $\mathcal{N}(71, 6.25)$ .

• (a) What percent of American males are over 6 feet, 2 inches tall?

• (b) What percent of those over 6 feet tall are also over 6 feet, 5 inches tall?

#### Solution

### (a)

Let H be a random variable measuring the height of American males in inches, so  $H \sim \mathcal{N}(71, 2.5^2)$ . Thus  $H \sim 2.5Z + 71$ , and:

$$egin{aligned} P[H \ge 74] & \gg \gg & 1 - P[H \le 74] \\ & \gg \gg & 1 - P[2.5Z + 71 \le 74] \\ & \gg \gg & 1 - P[Z \le 1.20] \\ & \gg \gg & 1 - 0.8849 pprox 11.5\% \end{aligned}$$

(b)

We seek  $P[H \geq 77 \mid H \geq 72]$  as the answer. Compute as follows:

$$\begin{split} P[H \ge 77 \mid H \ge 72] &= \frac{P[H \ge 77]}{P[H \ge 72]} \\ & \gg \gg \frac{P[2.5Z + 71 \ge 77]}{P[2.5Z + 71 \ge 72]} \\ & \gg \gg \frac{1 - P[Z \le 2.4]}{1 - P[Z \le 0.4]} = \frac{1 - 0.9918}{1 - 0.6554} \approx 2.38\% \end{split}$$

 $\equiv$  Example - Variance of normal from CDF table

Suppose  $X \sim \mathcal{N}(5, \sigma^2)$ , and suppose you know P[X > 9] = 0.2.

Find the approximate value of  $\sigma^2$  using a  $\Phi$  table.

## Solution

 $X \sim \mathcal{N}(5,\sigma^2) ext{ implies } X \sim \sigma Z + 5.$ 

So  $1 - P[X \le 9] = 0.2$  and thus  $P[\sigma Z + 5 \le 9] = 0.8$ . Then:

$$P[\sigma Z+5\leq 9] \quad = \quad P[Z\leq 4/\sigma]$$

so  $P[Z \leq 4/\sigma] = 0.8.$ 

Looking in the chart of  $\Phi$  for the nearest inverse of 0.8, we obtain  $4/\sigma = 0.842$ , hence  $\sigma = 4.75$ .