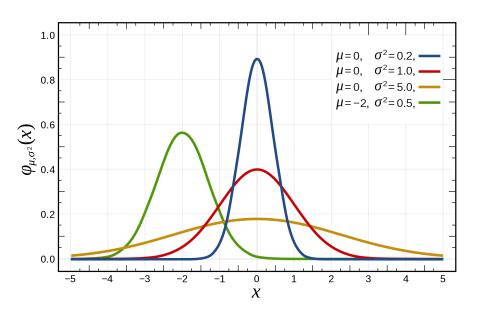
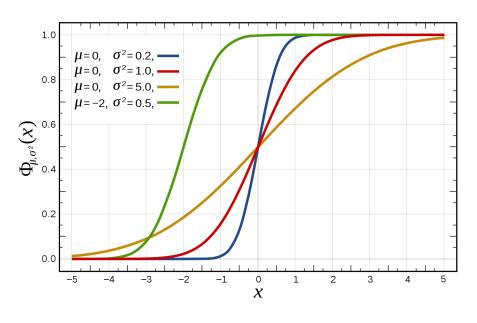
#### This Lecture ...

- Gaussian Random variables
- ► Monotone functions of continuous random variables
- Right-continuity of CDF
- Mixed random variables

# Gaussian random variable $(\mathcal{N}(\mu, \sigma^2))$

- ightharpoonup This is a real valued r.v. with two parameters,  $\mu$  and  $\sigma$ .
- Its pdf  $f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2}(\frac{x-\mu}{\sigma})^2}$  for all  $x \in \mathbb{R}$ .
- ▶ Verify:  $\int_{-\infty}^{\infty} f_X(x) dx = 1$ ,  $E[X] = \mu$  and  $Var(X) = \sigma^2$ .





# Standard Normal random variable $(\mathcal{N}(0,1))$

- ▶ When  $\mu = 0$  and  $\sigma = 1$ , it is called as a standard normal.
- ► In this case  $f_X(x) = \frac{1}{\sqrt{2\pi}} e^{\frac{-x^2}{2}}$ .
- What is  $\int_{-\infty}^{\infty} e^{\frac{-x^2}{2}} dx$ ? How do you even solve this? (=  $\sqrt{2\pi}$ )
- ▶ The CDF of standard normal, denoted by  $\Phi(x)$  is given by

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

- $ightharpoonup \Phi$  =These values are recorded in a table. (Fig. 3.10 in Bertsekas)
- https://en.wikipedia.org/wiki/Gaussian\_function

## Normality preserved under Linear Transformations

If  $X \sim \mathcal{N}(\mu, \sigma^2)$ , then Y = aX + b is also a normal variable with  $E[Y] = a\mu + b$  and variance  $a^2\sigma^2$ . (To be proved later)

## Significance of Gaussian r.v.

- Key role in Central limit theorem.
- ▶  $\frac{1}{n} \sum_{i=1}^{n} X_i \sim \mathcal{N}(\mu, \frac{\sigma^2}{n})$  where  $X_i$  is any random variable with mean  $\mu$  and variance  $\sigma^2$ .
- Building block for multinomial Gaussian vector and Gaussian processes (GP).
- Gaussian process are used in Bayesian Optimization (black-box optimization).
- Brownian motion is a type of GP and is used in Finance.
- ► GP Regression, Gaussian mixture models, used widely in ML.

## List of Probability distributions ...

https://en.wikipedia.org/wiki/List\_of\_probability\_distributions

Important ones are Beta, Gamma, Erlang, Logistic, Weibull ....

- Consider Y = aX + b where X is a continuous random variable.
- ightharpoonup What is  $F_Y(y)$  and  $f_Y(y)$ ?
- $F_Y(y) = P(Y \le y) = P(aX + b \le y).$
- $ightharpoonup F_Y(y) = F_X(\frac{y-b}{a}) \text{ if } a>0$
- $F_Y(y) = \frac{dF_Y(y)}{dy} = \frac{1}{a} f_X(\frac{y-b}{a}) \text{ when } a > 0$
- $F_Y(y) = 1 F_X(\frac{y-b}{a}) \text{ if } a < 0$
- $F_Y(y) = \frac{dF_Y(y)}{dy} = \frac{-1}{a} f_X(\frac{y-b}{a}) \text{ when } a < 0$
- ▶ In general,  $f_Y(y) = \frac{1}{|a|} f_X(\frac{y-b}{a})$

Consider Y = aX + b where X is a continuous random variable. Then  $f_Y(y) = \frac{1}{|a|} f_X(\frac{y-b}{a})$ .

- ▶ What if Y = g(X) where  $g(\cdot)$  is continuous, differentiable and monotone. Any guess?
- Since g(.) is monotone and continuous it is invertible. Let h(.) denote the inverse function. Then h(Y) = X.

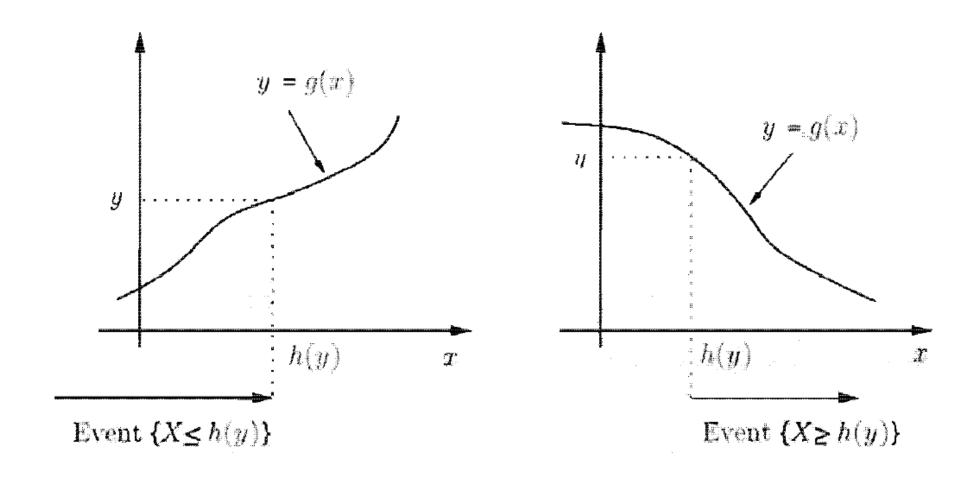
Consider Y = g(X) where g is monotone, continuous, differentiable. Then  $f_Y(y) = \left| \frac{dh}{dy}(y) \right| f_X(h(y))$  where h is the inverse function of g.

Consider Y = g(X) where g is monotone, continuous, differentiable. Then  $f_Y(y) = f_X(h(y)) \left| \frac{dh}{dy}(y) \right|$  where h is the inverse function of g.

#### Proof:

- Since g(.) is monotone and continuous it is invertible. Let h(.) denote the inverse function. Then X = h(Y).
- $F_Y(y) = P(g(X) \le y).$
- ▶ Is  $P(g(X) \le y) = P(X \le h(y))$  always?
- ▶ Are the two events  $\{g(X) \le y\}$  and  $X \le h(y)$  same?
- If they are same, then the two probabilities are equal.

▶ Are the two events  $\{g(X) \le y\}$  and  $\{X \le h(y)\}$  same ?



Same when g is increasing and compliments when g is decreasing.

- ▶ Are the two events  $\{g(X) \le y\}$  and  $\{X \le h(y)\}$  same ?
- $\triangleright$  Same when g is increasing and compliments when g is decreasing.
- ightharpoonup CASE 1: g(x) is non-decreasing
- $F_Y(y) = P(g(X) \le y) = P(X \le h(y)) = F_X(h(y)).$
- ►  $f_Y(y) = \frac{d}{dy}(F_X(h(y))) = f_X(h(y))\frac{dh}{dy}(y)$  where  $\frac{dh}{dy}(y) \ge 0$  as h is also non-decreasing.
- ▶ Rewritten therefore as  $f_Y(y) = f_X(h(y)) |\frac{dh}{dy}(y)|$

- ▶ Are the two events  $\{g(X) \le y\}$  and  $\{X \le h(y)\}$  same ?
- $\triangleright$  Same when g is increasing and compliments when g is decreasing.
- ightharpoonup CASE 2: g(x) is non-increasing
- $ightharpoonup F_Y(y) = P(g(X) \le y) = P(X > h(y)) = 1 F_X(h(y)).$
- ►  $f_Y(y) = -\frac{d}{dy}(F_X(h(y))) = -f_X(h(y))\frac{dh}{dy}(y)$  where  $\frac{dh}{dy}(y) \le 0$  as h is non-increasing as well.
- ▶ Rewritten therefore as  $f_Y(y) = f_X(h(y)) |\frac{dh}{dy}(y)|$ .

HW: What about the case when g is not monotone? Q: Suppose  $Y = X^2$ , then what is  $f_Y(y)$  in terms of  $f_X(x)$ ?