

Weds: HW by 5pm

- note problem that requires random numbers.

Weds: Read 4.3

Probabilities: Continuous Outcomes

Suppose the events / outcomes are labeled by a continuous variable, x . Then the probability density satisfies:

$$\boxed{\begin{aligned} 0 &\leq P(x) \\ \int_{-\infty}^{\infty} P(x) dx &= 1 \end{aligned}}$$

— positive
— normalized

The mean, or expectation value of the distribution is:

$$\boxed{\bar{x} \equiv \langle x \rangle = \int_{-\infty}^{\infty} x P(x) dx}$$

In many cases this quantifies the typical value returned when the distribution is sampled. The range of values returned is quantified via the standard deviation

$$\boxed{\sigma_x = \sqrt{\overline{x^2} - (\bar{x})^2}}$$

This requires calculation of

$$\overline{x^2} = \int_{-\infty}^{\infty} x^2 P(x) dx$$

Example: Consider the distribution:

$$P(x) = \begin{cases} 0 & x \leq 0 \\ \frac{2}{L^2}x & 0 \leq x \leq L \\ 0 & L \leq x \end{cases}$$

Determine:

a) mean (expectation value)

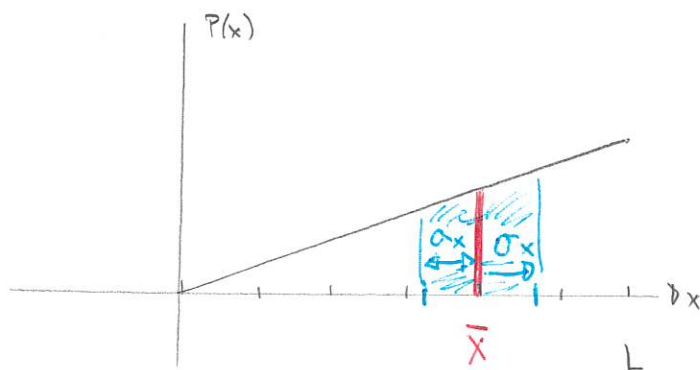
b) standard deviation

Answer a) $\bar{x} = \int_{-\infty}^{\infty} x P(x) dx$

$$\bar{x} = \int_0^L x \left(\frac{2}{L^2} x \right) dx$$

$$= \frac{2}{L^2} \int_0^L x^2 dx$$

$$= \frac{2}{L^2} \left. \frac{x^3}{3} \right|_0^L \Rightarrow \bar{x} = \frac{2}{3}L$$



b) use $\sigma_x = \sqrt{\overline{x^2} - (\bar{x})^2}$ Then

$$\overline{x^2} = \int_{-\infty}^{\infty} x^2 P(x) dx = \int_0^L x^2 \frac{2}{L^2} x dx = \frac{2}{L^2} \int_0^L x^3 dx$$

$$\overline{x^2} = \frac{1}{2L^2} L^4 = \frac{L^2}{2}$$

$$\text{Thus } \sigma_x = \sqrt{\frac{L^2}{2} - \left(\frac{2}{3}L\right)^2} = \sqrt{L^2 \left(\frac{1}{2} - \frac{4}{9}\right)} = L \sqrt{\frac{1}{18}}$$

$$\Rightarrow \sigma_x = \frac{1}{\sqrt{2}} \frac{L}{3}$$

Gaussian or normal distribution

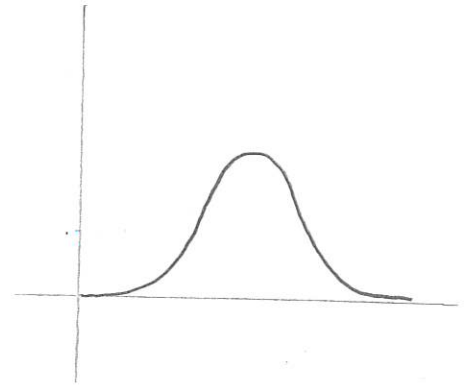
We often want to consider situations where the likely outcomes are concentrated in a narrow region amongst all the possibilities.

Demo: Plinko (PhET)

-use 26 Rows -run- show distribution

We note that the distribution is peaked around the midpoint (for equal left/right probability). If we tilt the left/right probability the peak shifts. The particular distribution in this case is called the binomial distribution.

In the limit as the number of levels approaches ∞ the distribution can be approximated by a Gaussian, or normal distribution:



$$P(x) = B e^{-(x-x_0)^2/2a^2}$$

Quiz 1 40%

We can fix the constant B by normalizing. We need

$$\int_{-\infty}^{\infty} P(x) dx = 1 \quad \Rightarrow \quad B \int_{-\infty}^{\infty} e^{-(x-x_0)^2/2a^2} dx = 1$$

Then integrals of this type can be evaluated.

Slide

In general, we have that:

$$\int_0^{\infty} e^{-\alpha x^2} dx = \frac{1}{2} \sqrt{\frac{\pi}{\alpha}}$$

$$\int_{-\infty}^{\infty} e^{-\alpha x^2} dx = \sqrt{\frac{\pi}{\alpha}}$$

$$\int_{-\infty}^{\infty} e^{-(\alpha x^2 + \beta x + \gamma)} dx = \sqrt{\frac{\pi}{\alpha}} e^{-(\beta^2 - 4\alpha\gamma)/4\alpha}$$

In this situation we use $u = x - x_0 \Rightarrow du = dx$

$$\Rightarrow B \int_{-\infty}^{\infty} e^{-u^2/2a^2} du = 1$$

$$\Rightarrow B \sqrt{\frac{\pi}{1/2a^2}} = 1 \Rightarrow B \sqrt{2\pi a^2} = 1$$

Assuming $a > 0$ we get $B = \frac{1}{\sqrt{2\pi}} \frac{1}{a}$

Thus a general form of the normal distribution is:

$$P(x) = \frac{1}{\sqrt{2\pi}} \frac{1}{a} e^{-(x-x_0)^2/2a^2}$$

We can compute the mean of this distribution. Then

$$\bar{x} = \int_{-\infty}^{\infty} x P(x) dx = \frac{1}{\sqrt{2\pi}} \frac{1}{a} \int_{-\infty}^{\infty} x e^{-(x-x_0)^2/2a^2} dx$$

look up / MAPLE

$$= x_0$$

Thus we have

$$P(x) = \frac{1}{\sqrt{2\pi}} \frac{1}{a} e^{-(x-\bar{x})^2/2a^2}$$

Quiz 2 - 10%

The width of the distribution is determined by a alone

Slide 2,3

We can show that

The distribution $P(x) = \frac{1}{\sqrt{2\pi}} \frac{1}{a} e^{-(x-\bar{x})^2/2a^2}$ the standard deviation is $\sigma = a$

Proof: We have $\sigma = \sqrt{\overline{x^2} - \bar{x}^2}$ and

$$\begin{aligned}\overline{x^2} &= \int_{-\infty}^{\infty} x^2 P(x) dx = \frac{1}{\sqrt{2\pi}} \frac{1}{a} \int_{-\infty}^{\infty} x^2 e^{-(x-\bar{x})^2/2a^2} dx \\ &= \frac{1}{\sqrt{2\pi}} \frac{1}{a} \sqrt{2\pi} a (a^2 + \bar{x}^2) = a^2 + \bar{x}^2\end{aligned}$$

$$\text{Then } \overline{x^2} - \bar{x}^2 = a^2 \Rightarrow \sigma = a \quad \square$$

The normal distribution is important in probability theory in general because of the central limit theorem. This says that if:

- 1) one samples a distribution and calculates the sample average and standard deviation of the mean, then
- 2) the sample average will be distributed according to a normal distribution with true mean equal to the mean of the original distribution and with standard deviation equal to the standard deviation of the mean of the original distribution