

Lec 3: Some Probability

1. Probability Axioms (Kolmogorov)

set S "sample space"subsets A, B, \dots every subset A $P(A) \geq 0$ disjoint $A \cap B = \emptyset$, $P(A \cup B) = P(A) + P(B)$

$$P(S) = 1$$

conditional P of A given B $P(A|B)$

$$P(A|B) \equiv \frac{P(A \cap B)}{P(B)}$$

Bayes Theorem

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

law of total probability B disjoint with A_i and $\cup A_i = S$

$$P(B) = \sum_i P(B|A_i)P(A_i)$$

$$P(A) \approx$$

 $\frac{\# \text{ outcomes } A}{\text{total trials}}$ experimental
definition

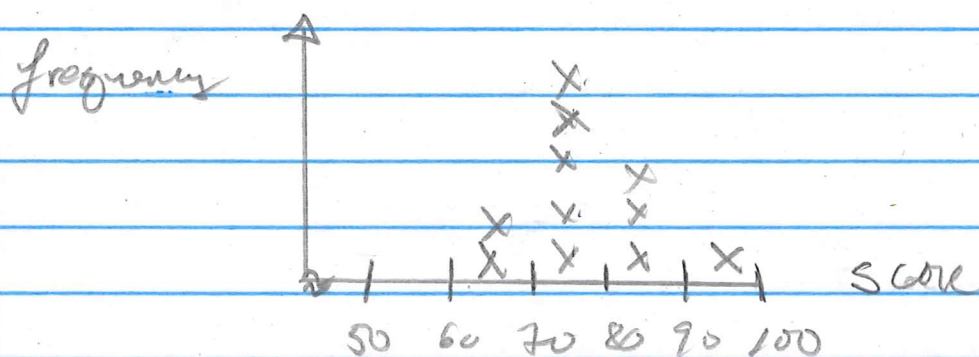
$$P(A+B) \leq P(A) + P(B)$$

equality if A, B disjoint

$$0 < P < 1$$

2. Histogram frequency distribution

consider exam grades of N students
grade distribution visualized by binning



bin	center	number
7	60.5	2
8	70.5	5
9	80.5	3
10	90.5	1

mean, expectation value

$$\langle x \rangle = \frac{1}{N} \sum N_i X_i$$

Variance

$$(\Delta x)^2 = \frac{1}{N} \sum N_i (X_i - \langle x \rangle)^2$$

Assume some underlying probability

$$P_i \approx \frac{N_i}{N} \quad \sum P_i = 1$$

$$\langle x \rangle = \sum P_i x_i$$

$$\langle x^2 \rangle = \sum P_i x_i^2$$

Variance $\langle (x - \langle x \rangle)^2 \rangle = \sum P_i (x_i - \langle x \rangle)^2$

$$= \sum P_i (x_i^2 - 2x_i \langle x \rangle + \langle x \rangle^2)$$

$$= \sum_i P_i x_i^2 - 2 \langle x \rangle \sum_i P_i x_i + \langle x \rangle^2 \sum P_i$$

$$= \langle x^2 \rangle - \langle x \rangle^2$$

3. Continuous Limit

probability density function

$$P_i \rightarrow P(x) dx \quad \text{PDF}$$

$$\int P(x) dx = 1$$

$$\langle x \rangle = \int P(x) x dx$$

Example: exponential decay

$$\frac{dN}{dt} = -\frac{N(t)}{\tau} \Rightarrow N(t) = N(0) e^{-t/\tau}$$

normalized PDF

$$P(t) = 0 \quad t < 0$$

$$P(t) = \frac{1}{\tau} e^{-t/\tau} \quad t \geq 0$$

lec 3-4

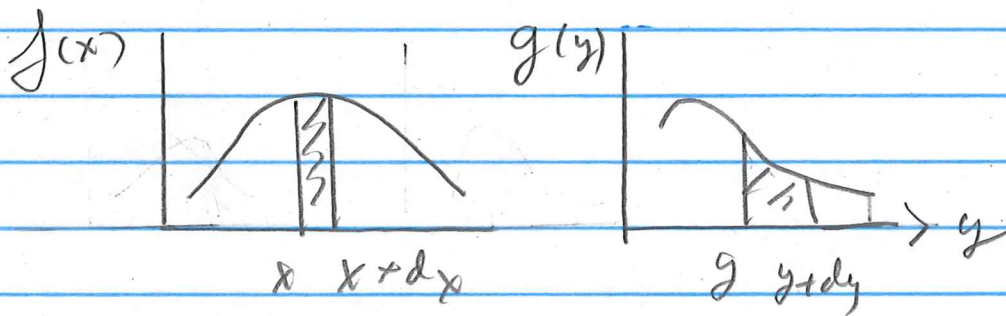
$$\langle t \rangle = \frac{1}{\tau} \int_0^{\infty} e^{-t/\tau} t dt = \tau \int_0^{\infty} e^{-x} x dx = \tau$$

useful integral: $\int_0^{\infty} x^n e^{-x} dx = n!$

$$\begin{aligned} \text{variance } \langle t^2 \rangle &= \frac{1}{\tau} \int_0^{\infty} e^{-t/\tau} (t-\tau)^2 dt \\ &= \tau^2 \int_0^{\infty} e^{-x} (x-1)^2 dx = \tau^2 [2! - 2 + 1] = \tau^2 \end{aligned}$$

$$\sigma_t = \tau$$

4 Change of variable $y = y(x)$
invert $x = x(y)$



$$f(x) dx = g(y) dy$$

$$g(y) dy = \left| \frac{dy}{dx} \right|^{-1} dy f(x(y))$$

PDF in y is $\left| \frac{dy}{dx} \right|^{-1} f(x(y))$

example $f(x) = e^{-x}$

$$y = e^{-x}$$

$$x(y) = -\ln y$$

$$\left| \frac{dy}{dx} \right| = e^{-x} = y$$

$$g(y) = \left| \frac{dy}{dx} \right|^{-1} f(x(y))$$

$$= \frac{1}{y} e^{\ln y} = 1 \quad \text{flat}$$

range $x [0, \infty]$ $y [0, 1]$

5 Correlation: two random variables x_1, x_2
and quantity $g(x_1, x_2)$

$$\int P(x_1, x_2) dx_1 dx_2 = 1$$

$$\langle g \rangle = \int P(x_1, x_2) g(x_1, x_2) dx_1 dx_2$$

expand g about $\langle x_1 \rangle, \langle x_2 \rangle$

$$g(x_1, x_2) = g(\langle x_1 \rangle, \langle x_2 \rangle) + \frac{\partial g}{\partial x_1} \Big|_{(\langle x_1 \rangle, \langle x_2 \rangle)} (x_1 - \langle x_1 \rangle)$$

$$+ \frac{\partial g}{\partial x_2} \Big|_{(\langle x_1 \rangle, \langle x_2 \rangle)} (x_2 - \langle x_2 \rangle)$$

$$\text{Then } \langle g \rangle = g(\langle x_1 \rangle, \langle x_2 \rangle)$$

$$(\Delta g)^2 = \langle (g - \langle g \rangle)^2 \rangle$$

$$= \left(\frac{\partial g}{\partial x_1} \right)^2 \langle (x_1 - \langle x_1 \rangle)^2 \rangle + \left(\frac{\partial g}{\partial x_2} \right)^2 \langle (x_2 - \langle x_2 \rangle)^2 \rangle$$

$$= 2 \frac{\partial g}{\partial x_1} \left| \frac{\partial g}{\partial x_2} \right| \langle (x_1 - \langle x_1 \rangle)(x_2 - \langle x_2 \rangle) \rangle$$

$\underbrace{\hspace{10em}}_{\text{Cov}(x_1, x_2) \text{ Covariance}}$

giving

$$(\Delta g)^2 = \left(\frac{\partial g}{\partial x_1} \right)^2 \Delta x_1^2 + \left(\frac{\partial g}{\partial x_2} \right)^2 \Delta x_2^2$$

$$+ 2 \frac{\partial g}{\partial x_1} \left| \frac{\partial g}{\partial x_2} \right| \text{Cov}(x_1, x_2)$$

if x_1, x_2 are uncorrelated

$$P(x_1, x_2) = P_1(x_1) P_2(x_2)$$

and $\text{Cov}(x_1, x_2) = 0$

linear correlation coefficient $-1 \leq r \leq 1$

$$r = \frac{\langle x_1 x_2 \rangle - \langle x_1 \rangle \langle x_2 \rangle}{\Delta x_1 \Delta x_2} = \frac{\text{Cov}(x_1, x_2)}{\Delta x_1 \Delta x_2}$$