

#2: Coin 1: $P(H) = 0.01$ } Choose a coin
 Coin 2: $P(H) = 0.99$ } uniformly, and flip twice

$$H_1 = P(\text{flip 1 is H})$$

$$H_2 = P(\text{flip 2 is H})$$

Are H_1 and H_2 ind?

$$P(H_1) = P(H_2) = \frac{1}{2}$$

$$P(H_2 | H_1) \approx 0.98$$

$P(H_2 | H_1) \neq P(H_2)$ so dependent.

#5: X_1, X_2, \dots, X_n are i.i.d. r.v.'s

(independent, identically distributed).

$$E[X_i] = \mu, \text{Var}(X_i) = \sigma^2$$

$$M_n = \frac{1}{n} \sum_{i=1}^n X_i$$

$$E[M_n] = E\left[\frac{1}{n} \sum_{i=1}^n X_i\right] = \frac{1}{n} E\left[\sum_{i=1}^n X_i\right]$$

$$= \frac{1}{n} \sum_{i=1}^n E[X_i] = \frac{1}{n} \sum_{i=1}^n \mu = \frac{1}{n} \cdot n\mu = \mu$$

$$\text{Var}(M_n) = \text{Var}\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n^2} \text{Var}\left(\sum_{i=1}^n X_i\right)$$

$$= \frac{1}{n^2} \sum_{i=1}^n \text{Var}(X_i)$$

because X_i 's ind

$$= \frac{1}{n^2} \sum_{i=1}^n \sigma^2 = \frac{1}{n^2} \cdot n\sigma^2 = \frac{\sigma^2}{n}$$

Continuous Random Variables (slide pack 7)

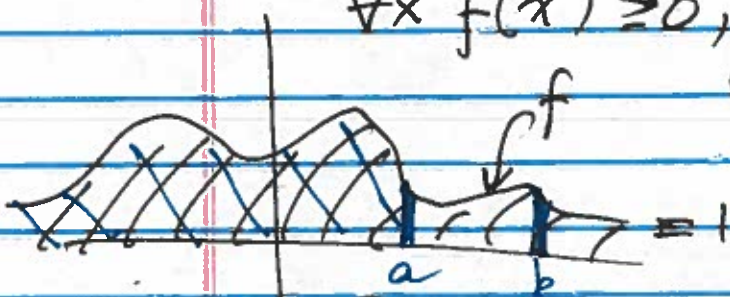
A continuous r.v. takes values from an uncountable set. (For us, either the real numbers \mathbb{R} or some interval of reals.)

Ex. height of a randomly chosen adult
time until next packet arrives at a server

Defn. $f: \mathbb{R} \rightarrow \mathbb{R}$ is a probability density function (pdf) (or density) for a continuous random variable X iff

$$\forall x \ f(x) \geq 0, \text{ and } \int_{-\infty}^{+\infty} f(x) dx = 1.$$

(normalized, just like a pmf, where $\sum_x p(x) = 1$.)



Defn. The cumulative distribution function (cdf)

$F: \mathbb{R} \rightarrow \mathbb{R}$ associated with pdf f is

$$F(a) = P(X \leq a) = \int_{-\infty}^a f(x) dx$$

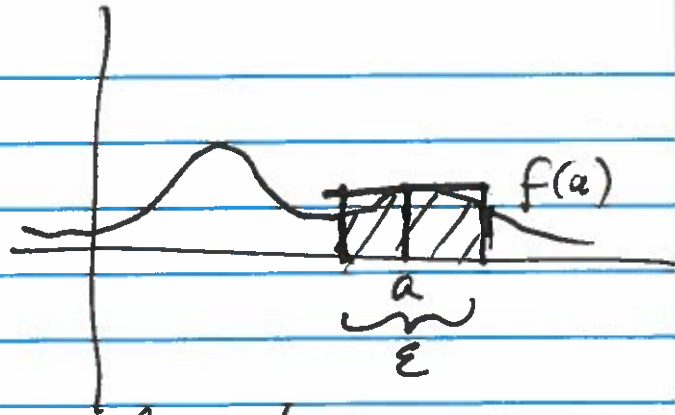
Note: $P(a < X \leq b) = F(b) - F(a) = \int_a^b f(x) dx$

Note: $f(x) = \frac{d}{dx} F(x)$

Densities are not probabilities: they may be greater than 1.



$$\begin{aligned}
 P\left(a - \frac{\epsilon}{2} < X \leq a + \frac{\epsilon}{2}\right) \\
 &= \int_{a - \frac{\epsilon}{2}}^{a + \frac{\epsilon}{2}} f(x) dx \\
 &\approx \epsilon f(a)
 \end{aligned}$$



The probability that X is close to a is proportional to $f(a)$, and this approximation improves as ϵ gets smaller. So at $\epsilon = 0$, this says $P(X = a) = 0$.

For continuous r.v.'s, usually substitute \int for \sum :

$$E[X] = \int_{-\infty}^{+\infty} x f(x) dx$$