

Mean

$\{x\}$ — dataset
 x_i — i 'th item

N items

$$\text{mean}(\{x\}) = \frac{1}{N} \sum_{i=1}^N x_i$$

- scaling data scales the mean
 $\text{mean}(\{kx\}) = k \text{mean}(\{x\})$
- translating data translates the mean
 $\text{mean}(\{x+c\}) = \text{mean}(\{x\}) + c$
- $\sum_{i=1}^N (x_i - \text{mean}(\{x\})) = 0$

Choose μ such that the sum of squared dist.'s from data points to μ is minimized

$$\mu = \underset{a}{\operatorname{argmin}} \sum_{i=1}^N (x_i - \mu)^2$$

$$\mu = \operatorname{mean}(\{x_i\})$$

$$\frac{d}{d\mu} \sum_{i=1}^N (x_i - \mu)^2 = 2 \sum_{i=1}^N (x_i - \mu) = 0$$

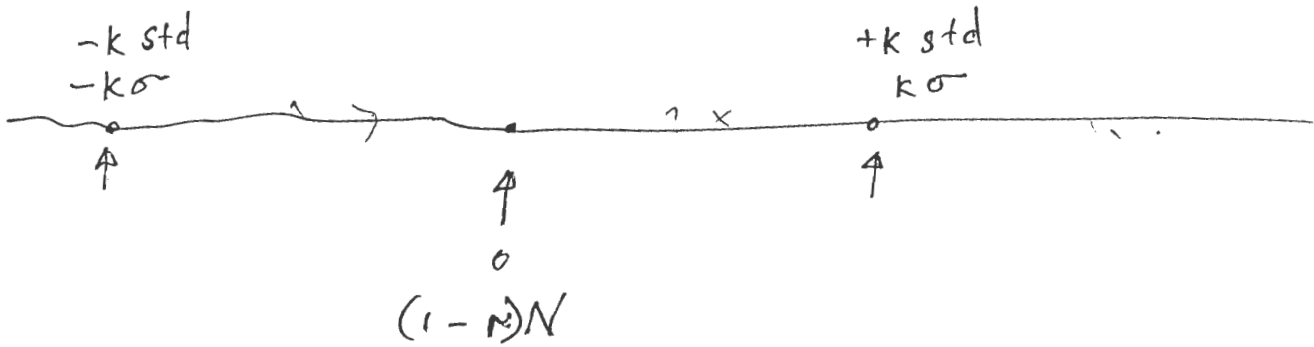
$$\sum_{i=1}^N (x_i - \mu) = \sum_{i=1}^N x_i - N\mu = 0$$
$$\mu = \frac{1}{N} \sum_{i=1}^N x_i$$

$$\text{std}(\{x\}) = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \text{mean}(\{x\}))^2}$$

$$= \sqrt{\text{mean}(\{(x_i - \text{mean}(\{x\}))^2\})}$$

- translating data does not change std
 $\text{std}(\{x+c\}) = \text{std}(\{x\})$
- scaling data scales std.
 $\text{std}(\{kx\}) = k \text{std}(\{x\})$
- for any dataset there are at most $\frac{N}{k^2}$ data items k std away from the mean.
- & there is at least 1 data item 1 std away from the mean

k std away from mean.



$N(1-r)N$ ← at k std away from

$$\sigma = \sqrt{\frac{(rN)k^2\sigma^2}{N}}$$

$$1 = \sqrt{rk^2} \quad \text{so} \quad r = \frac{1}{k^2}$$

- $\approx 68\%$ within 1 std.
- $\approx 95\%$ within 2 std.
- $\approx 99\%$ within 3 std.

$$\text{std}(\{x_i\})^2 \leq \max_i (x_i - \text{mean}(\{x_i\}))^2$$

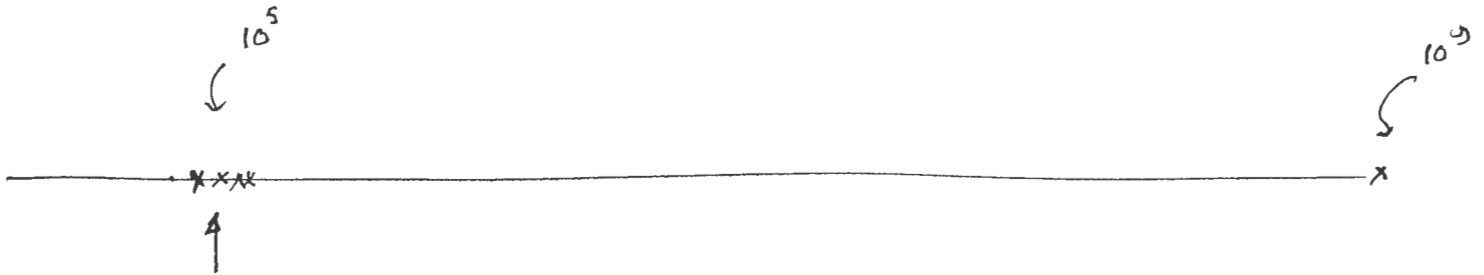
$$\cancel{\text{std}(\{x_i\})^2} = \cancel{\frac{1}{N} \sum_i (x_i - \text{mean}(\{x_i\}))^2} \leq \max_i (x_i - \text{mean}(\{x_i\}))^2$$

$$\text{var}(\{x_i\}) = \text{std}(\{x_i\})^2$$

$$\frac{10 \cdot 10^5 + 10^9}{11}$$

median $(\{x\}) =$

order the data
take the datapoint
in the middle
of their ordering



Interquartile range:

%

25%ile

75%ile

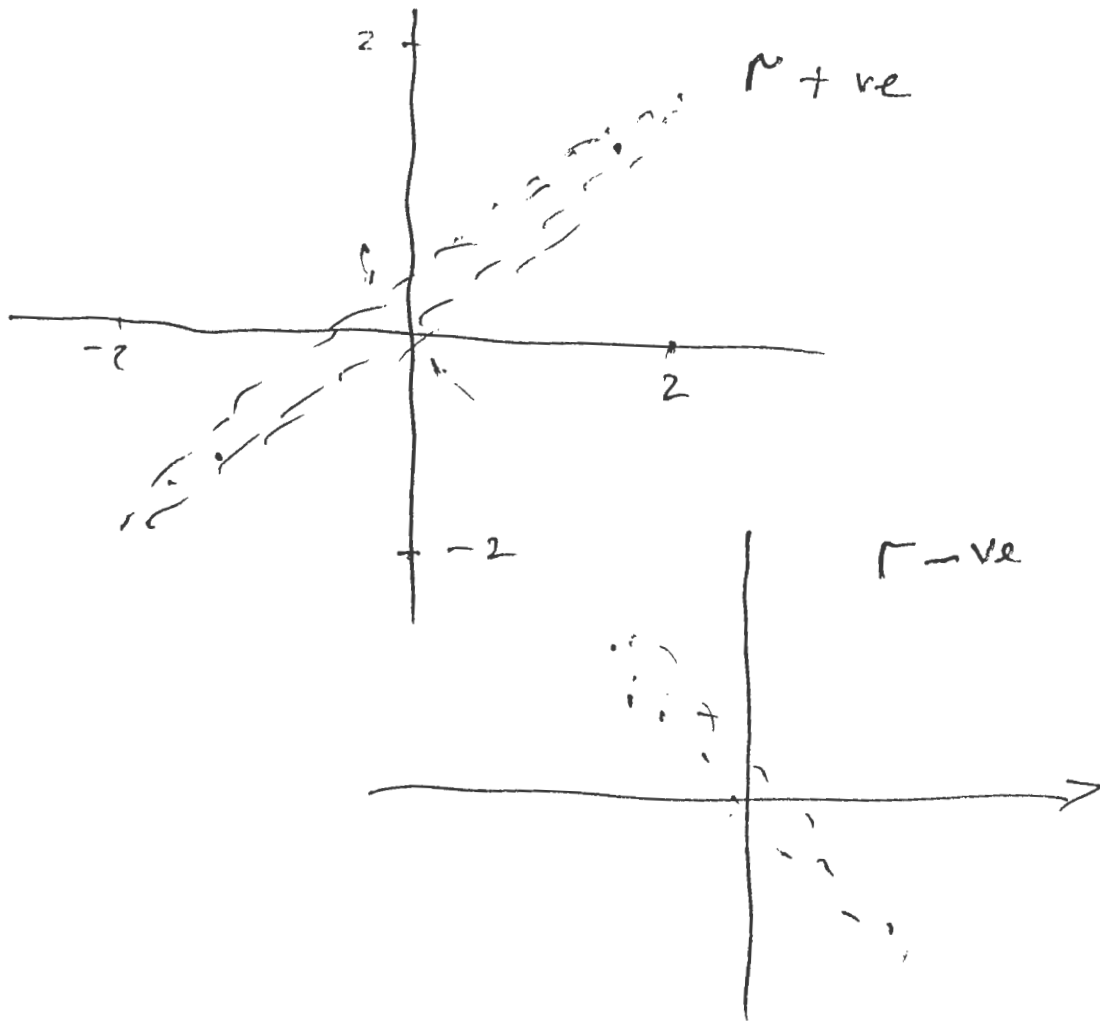
$$\text{mean}(\hat{x}) = 0$$

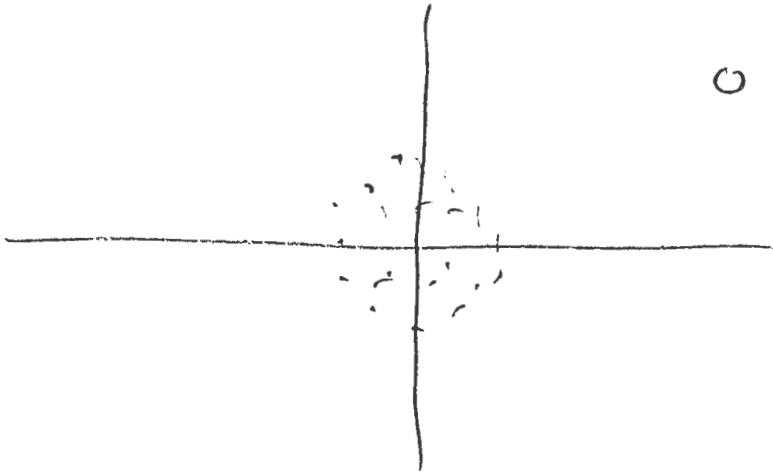
$$\text{mean}(\hat{y}) = 0$$

$$\text{mean}(\hat{x}^2) = 1$$

$$\text{mean}(\hat{y}^2) = 1$$

$$\text{Mean}(\hat{x}\hat{y}) = r = \text{correlation coefficient}$$





o

$$1) \text{corr}(x, y) = r = \text{corr}(y, x)$$

$$2) \text{corr}(\{ax+b, cy+d\}) \\ = \text{sign}(ac) \text{corr}(x, y)$$

$$\# -) \leq \text{corr}(x, y) \leq 1$$

$$\text{mean}(\hat{x} \hat{y}) = \frac{1}{N} \sum_{i=1}^N \hat{x}_i \cdot \hat{y}_i = \underline{x} \cdot \underline{y}$$

$$\frac{1}{\sqrt{N}} \begin{pmatrix} \hat{x}_1 \\ \vdots \\ \hat{x}_N \end{pmatrix} = \underline{x} \quad \underline{y} = \frac{1}{\sqrt{N}} \begin{pmatrix} \hat{y}_1 \\ \vdots \\ \hat{y}_N \end{pmatrix} \quad \underline{\underline{y}}$$
$$\underline{x} \cdot \underline{x} = 1 \quad \underline{y} \cdot \underline{y} = 1$$

(\hat{x}_i, \hat{y}_i) , N etc

$(\hat{x}_0, ?)$ y^p

$$\hat{y}^p = a \hat{x}_0 + b \quad \leftarrow \quad b=0$$

$$u_i = \hat{y}_i - \hat{y}_i^p(x_i)$$

$$\text{mean}(u) = 0$$

$$\begin{aligned} \text{mean}(u) &= \text{mean}(\hat{y}_i - a \hat{x}_i - b) \\ &= \text{mean}(\hat{y}_i) - a \text{mean}(\hat{x}_i) - b \end{aligned}$$

$$\Rightarrow b=0$$

minimize $(\text{var}(u))$

$$\text{var}(u) = \text{var}(\hat{y}_i - a \hat{x}_i)$$

$$\begin{aligned} &= \text{mean}((\hat{y}_i - a \hat{x}_i)^2) \\ &= \text{mean}(\hat{y}_i^2) - 2a \text{mean}(\hat{x}_i \hat{y}_i) \\ &\quad + a^2 \text{mean}(\hat{x}_i^2) \end{aligned}$$

$$= 1 - 2ar + a^2$$

$$\frac{d(\text{var}(u))}{da} = -2r + 2a = 0$$

FACT:

$$\hat{x}_0, ? \rightarrow \hat{x}_0, r x_0$$

$$?, \hat{y}_0 \rightarrow r \hat{y}_0, \hat{y}_0$$

$$?, \hat{y}_0$$

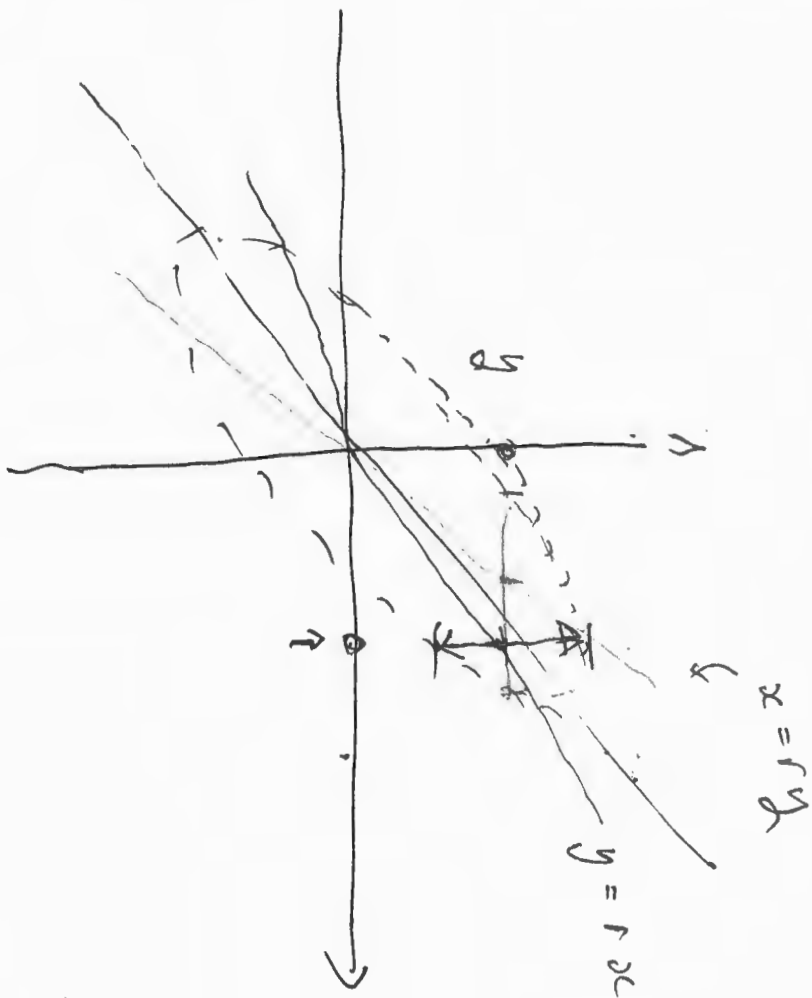
$$\hat{x}^p = a y + b$$

$$u_i = \hat{x}_i - \hat{x}_i^p$$

$$\text{mean}(u) = 0 = \text{mean}(\hat{x}) - a \text{mean}(\hat{y}) - b$$
$$\Rightarrow b = 0$$

$$\text{var}(u) = \text{mean} \left((\hat{x}_i - a \hat{y}_i)^2 \right)$$
$$= \text{mean}(\hat{x}_i^2) - 2a \text{mean}(\hat{x}_i \hat{y}_i) + a^2 \text{mean}(\hat{y}_i^2)$$
$$= 1 - 2a r + a^2$$

$$\rightarrow a = r$$



$$u_i = \hat{y}_i - y_i$$

$$\begin{aligned} \text{mean}(u_i^2) &= \text{var}(u_i) = 1 - 2\rho r + \rho^2 \\ &= 1 - \rho^2 \end{aligned}$$

$$\text{std}(u_i) = \sqrt{1 - \rho^2}$$

1) move to N.C.s

$$\hat{x}_i = \frac{(x_i - \text{mean}(x))}{\text{std}(x)} \quad \hat{y}_i = \frac{(y_i - \text{mean}(y))}{\text{std}(y)}$$

2) get $r = \text{mean}(\hat{x}_i \hat{y}_i)$

3) predict $(\hat{x}_0, r \hat{x}_0)$

4) goback to non normalized coords

$$\begin{aligned} \overset{\text{non}}{\underset{\uparrow}{y}}^p &= \text{std}(y) \hat{y}^p + \text{mean}(y) \\ &= \text{std}(y) r \hat{x}_0 + \text{mean}(y) \\ &= \text{std}(y) r \frac{(x_0 - \text{mean}(x))}{\text{std}(x)} + \text{mean}(y) \end{aligned}$$

$\{H, T\}$ $\{1, 2, 3, 4, 5, 6\}$ $\{s, \bar{s}\}$ Ω $\{BG, GB, BBG, BBB, \\ GGB, GGG\}$ $\{GGG, GCC, GGC\}$ $\{CFM, CMF, FCM, \dots\}$ $\{K, Q, J\}$

$B^* G^+ B$

$P(A)$ meaning $\lim_{N \rightarrow \infty} \frac{\#(A)}{N}$

$$0 \leq P(A) \leq 1$$

$$\sum_{A_i \in \Omega} P(A_i) = 1$$

————— →

$\{1, 2, 3, 4, 5, 6\}$

Ω

$S = \{\text{all even rolls}\}$ ← Events

$P(S)$

E_1 event

E_2 event

$E_1 \cap E_2$ is event

$E_1 \cup E_2$ is event

E event $\Rightarrow \Omega - E$ event

Σ ← event space

$\emptyset \in \Sigma$, $\Omega \in \Sigma$

$U \in \Sigma, V \in \Sigma \Rightarrow U \cup V \in \Sigma$

$U \cap V \in \Sigma$

$U \in \Sigma \Rightarrow U^c \in \Sigma$

$P(X|Y)$ $P(Y|X)$

$$P(Y|X) = \frac{P(X|Y) P(Y)}{P(X)}$$

$X(\omega)$

$$P(\{\omega: X(\omega) = x\})$$

$$\uparrow \quad P(x)$$

$$P(\{\omega: X(\omega) \leq x\})$$

$$\{\omega: X(\omega) = x_1\} \cap \{\omega: X(\omega) = x_2\}$$

$$\{\omega: X(\omega) = x_1\} \cap \{\omega: X(\omega) = x_2\}$$

$$\sum_{x \in D} P(\{\omega: X(\omega) = x\}) = 1$$

$$\sum_{x \in D} P(x) = 1$$

1

1 with p
0 $(1-p)$

$P(0)$ $(1-p)^2$
 $P(1)$ $p(1-p)$
 $P(2)$ $p(1-p)$
 $P(3)$ p^2

$$P(H) = p$$

q with prob p
 $-N$ with $(1-p)$

X, Y

$$\{\omega: X(\omega) = x\}$$

$$\{\omega: Y(\omega) = y\}$$

$$P(\{\omega: X(\omega) = x\} \cap \{\omega: Y(\omega) = y\})$$

joint probability $P(x, y)$

$$P(\{X=x\} | \{Y=y\}) = \frac{P(\{X=x\} \cap \{Y=y\})}{P(\{Y=y\})}$$

$$P(x|y) = \frac{P(x, y)}{P(y)}$$

$$P(x|y) = \frac{P(y|x) P(x)}{P(y)}$$

$$\sum_{x \in D} P(x | y) = 1$$

A_i and A_j st $A_i \cap A_j = \emptyset \quad i \neq j$
 $\cup A_i = \Omega$

$$\sum_{y \in D_y} P(x, y) = \sum_{y \in D_y} P(\{\omega: X(\omega) = x\} \cap \{\omega: Y(\omega) = y\})$$

$= P(x)$



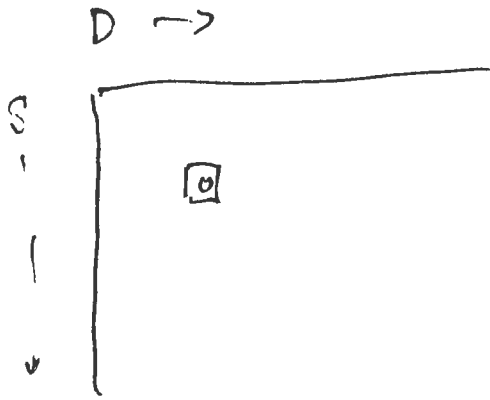
X, Y

$$S = X + Y$$

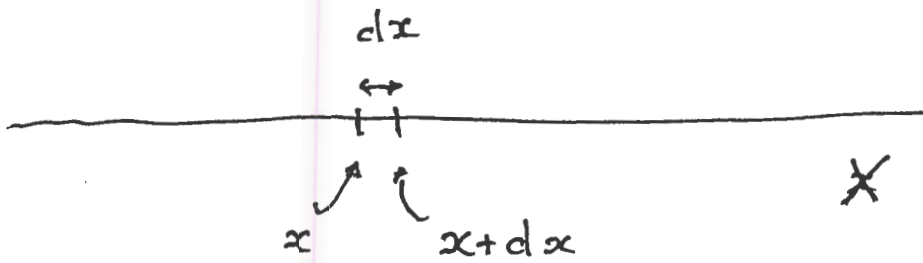
$$D = X - Y$$

$$S = 2$$

$$S = 3$$



$$P(x, y) = P(x) P(y)$$



$$P(\{\omega: X(\omega) \in [x, x+dx]\}) = p(x) dx$$

↑
probability
density fu

$$P(\{\omega: X(\omega) \in [a, b]\}) \leftarrow P(\{X=x\}) \leftarrow p(x)$$

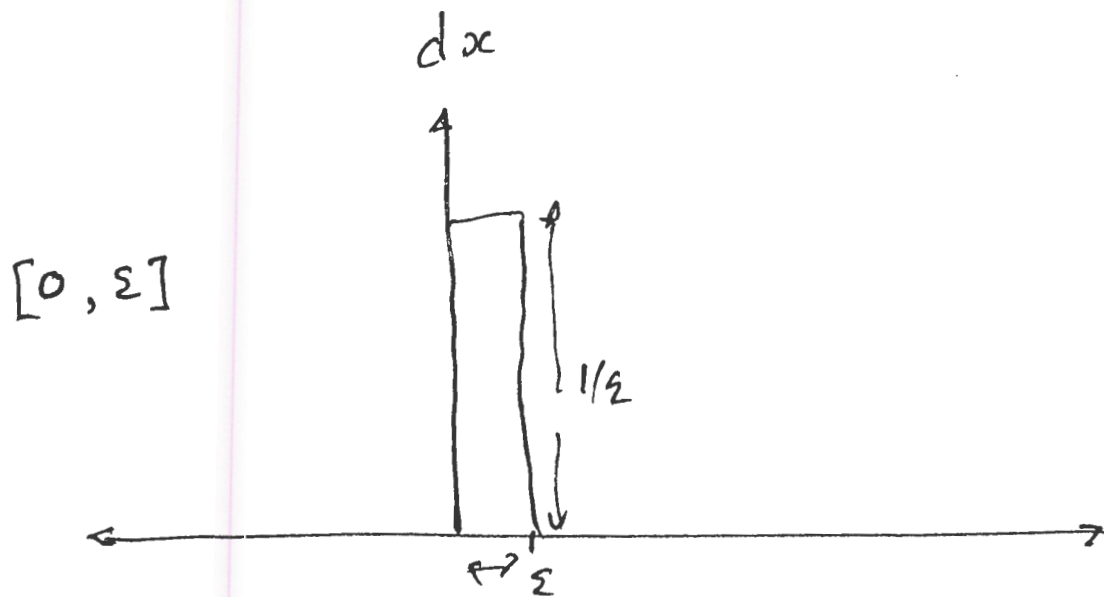
$$= \int_a^b p(x) dx.$$

$$\int_{-\infty}^{\infty} p(x) dx = 1$$

$$g(x) \geq 0$$

$$\frac{g(x)}{\int_{-\infty}^{\infty} g(x) dx} \leftarrow \text{Normalizing const.}$$

the prob that X is in $[x, x+dx]$



$$P(H) = p \quad ; \quad P(T) = 1-p$$

Heads, I get q
T, I give $r =$ get $-r$

N times

$$\bullet \frac{Npq + (-r)N(1-p)}{N}$$

$$pq + (1-p)(-r)$$

Expectation, Expected value

$$E(X) = \sum_{x \in D} x P(\{\omega: X(\omega) = x\})$$

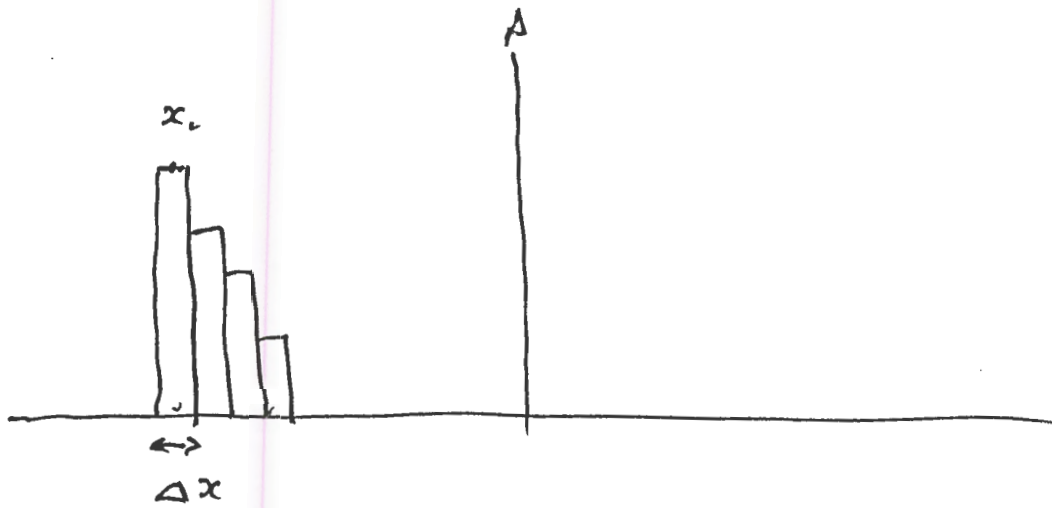
$$f : X \rightarrow D_f$$

New RV called F

$$\sum_{u \in D_f} u P(\{F = u\})$$

$$\sum_{x \in D} f(x) P(\{X = x\}) = E[f]$$

$p(x)$



$p(x_i) \Delta x$

$$\sum_{x_i} x_i p(x_i) \Delta x$$

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

$$\int_{-\infty}^{\infty} f(x) p(x) dx = E_{P_t}(f)$$

$$E[0]$$

$$E[kf] = kE[f]$$

$$E[f+g] = E[f] + E[g]$$

X

$$E[X]$$

← mean, expected value

$$E[(X - E[X])^2]$$

← variance

X, Y

$$E[(X - E[X])(Y - E[Y])]$$

← covariance