

## Probability Refresher

Let  $X$  and  $Y$  be random variables that take on values  $x_1, \dots, x_I$  and  $y_1, \dots, y_J$ , respectively, where  $x_i$  and  $y_j$  are real numbers.

$\Pr(x_i)$  = probability that  $X = x_i$

$\Pr(y_j)$  = probability that  $Y = y_j$

$$\therefore \sum_{i=1}^I \Pr(x_i) = 1$$

$$\varphi \sum_{j=1}^J \Pr(y_j) = 1$$

# Expected Value

Definition:

$$E(X) = \sum_{i=1}^I x_i \cdot Pr(x_i)$$

$$E(Y) = \sum_{j=1}^J y_j \cdot Pr(y_j)$$

Properties:

① For any constant,  $b$ ,

$$E(b + X) = b + E(X)$$

$$\neq E(b \cdot X) = b \cdot E(X)$$

Special case: using  $b = -E(X)$ ,

$$E[X - E(X)] = 0$$

②  $E(X + Y) = E(X) + E(Y)$

## Independence

Definition:  $X$  +  $Y$  are independent  
if  $\Pr(X, Y) = \Pr(X) \cdot \Pr(Y)$

Properties: If  $X$  +  $Y$  are independent,  
then

①  $X - a$  and  $Y - b$  are independent,  
for any constants,  $a$  and  $b$ .

Special Case:  $X - E(X)$  and  $Y - E(Y)$   
are independent

②  $E(X \cdot Y) = E(X) \cdot E(Y)$

Proof of Property ② :

$$E(X \cdot Y)$$

$$= \sum_{i,j} x_i \cdot y_j \cdot \text{Pr}(x_i, y_j)$$

$$= \sum_{i,j} x_i \cdot y_j \cdot \text{Pr}(x_i) \cdot \text{Pr}(y_j)$$

$$= \sum_{i,j} [x_i \cdot \text{Pr}(x_i)] \cdot [y_j \cdot \text{Pr}(y_j)]$$

$$= \left[ \sum_i x_i \cdot \text{Pr}(x_i) \right] \cdot \left[ \sum_j y_j \cdot \text{Pr}(y_j) \right]$$

$$= E(X) \cdot E(Y)$$

q.e.d.

## Variance

Definition:  $\text{Var}(X) = E[X - E(X)]^2$

Special Case: If  $E(X) = 0$  then

$$\text{Var}(X) = E(X^2)$$

Properties: for any constant,  $b$ ,

$$\textcircled{1} \text{Var}(X+b) = \text{Var}(X)$$

Special Case:  $\text{Var}[X - E(X)] = \text{Var}(X)$

$$\textcircled{2} \text{Var}(b \cdot X) = b^2 \cdot \text{Var}(X)$$

for any  $a$

Proof of Property ②:

$$\begin{aligned}\text{Var}(b \cdot X) &= E[bX - E(bX)]^2 \\ &= E[bX - b \cdot E(X)]^2 \\ &= E(b[X - E(X)])^2 \\ &= E(b^2 \cdot [X - E(X)]^2) \\ &= b^2 \cdot E[X - E(X)]^2 \\ &= b^2 \cdot \text{Var}(X)\end{aligned}$$

q.e.d.

Lemma 1:

If  $X + Y$  are independent

and  $E(X) = E(Y) = 0$

then  $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$

Proof:

$$\begin{aligned} & \text{Var}(X+Y) \\ &= E(X+Y)^2 \\ &= E(X^2 + Y^2 + 2XY) \\ &= E(X^2) + E(Y^2) + E(2XY) \\ &= \text{Var}(X) + \text{Var}(Y) + 2 \cdot E(X) \cdot E(Y) \end{aligned}$$

~~by Lemma 1~~

$$= \text{Var}(X) + \text{Var}(Y) + 2 \cdot 0 \cdot 0$$

$$= \text{Var}(X) + \text{Var}(Y)$$

~~QED~~  
QED

Theorem 2:

If  $X$  &  $Y$  are independent,  
then  $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$ .

Proof: Let  $X' = X - E(X)$  and  $Y' = Y - E(Y)$ .

$\therefore X'$  and  $Y'$  are independent, and

$$E(X') = E(Y') = 0$$

Also,

$$\text{Var}(X+Y)$$

$$= \text{Var}(X' + Y')$$

$$= \text{Var}(X') + \text{Var}(Y') \quad \text{by Lemma 1}$$

$$= \text{Var}(X) + \text{Var}(Y)$$

q.e.d.

Corollary 3:

If  $X_1, X_2, \dots, X_N$  are independent random variables, then

$$\text{Var}\left(\sum_i X_i\right) = \sum_i \text{Var}(X_i)$$

Proof: Similar to Theorem 2.

Corollary 4:

If  $X, X_1, X_2, \dots, X_N$  are independent and identically distributed (iid) random variables, then

$$\text{Var} \left( \sum_{i=1}^N X_i \right) = N \cdot \text{Var}(X)$$

Proof: Because  $X, X_1, \dots, X_N$  all have the same distribution,

$$\text{Var}(X) = \text{Var}(X_1) = \dots = \text{Var}(X_N)$$

$$\therefore \text{Var} \left( \sum_{i=1}^N X_i \right)$$

$$= \sum_{i=1}^N \text{Var}(X_i)$$

by Corollary 3

$$= \sum_{i=1}^N \text{Var}(X)$$

$$= N \cdot \text{Var}(X)$$

q ed.

Theorem 5 :

If  $X, X_1, X_2, \dots, X_N$  are independent and identically distributed (iid), and

$$\bar{X} = \frac{1}{N} \sum_{i=1}^N X_i$$

then 
$$\text{Var}(\bar{X}) = \frac{\text{Var}(X)}{N}$$

Proof:

$$\text{Var}(\bar{X})$$

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

$$= \frac{1}{N^2} \cdot \text{Var}\left(\sum_{i=1}^N X_i\right) \quad \text{by variance property } \textcircled{2}$$

$$= \frac{1}{N^2} \cdot N \cdot \text{Var}(X) \quad \text{by Corollary 4}$$

$$= \frac{1}{N} \cdot \text{Var}(X)$$

qed.