

COMP4804 Important Facts

Union of Events and Boole's Inequality. For any events A and B

$$\begin{aligned}\Pr\{A \text{ or } B\} &= \Pr\{A\} + \Pr\{B\} - \Pr\{A \text{ and } B\} \\ &\leq \Pr\{A\} + \Pr\{B\} .\end{aligned}$$

Conditional Probability.

$$\Pr\{A \mid B\} = \frac{\Pr\{A \text{ and } B\}}{\Pr\{B\}}$$

Another useful way of writing this is

$$\Pr\{A \text{ and } B\} = \Pr\{A \mid B\} \Pr\{B\} . \quad (1)$$

Independence. We say that A and B are *independent* if and only if

$$\Pr\{A \mid B\} = \Pr\{A\}$$

If A and B are independent then (1) becomes

$$\Pr\{A \text{ and } B\} = \Pr\{A\} \Pr\{B\} \quad (\text{Only if } A \text{ and } B \text{ are independent!})$$

Expected Value. For a random variable X

$$E[X] = \sum_x x \Pr\{X = x\} .$$

Linearity of Expectation. For any random variables X and Y

$$E[X + Y] = E[X] + E[Y] .$$

More generally, for any random variables X_1, \dots, X_n

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

Linearity of expectation, in combination with *indicator variables*, is extremely useful for things we can count.

Markov's Inequality. For any *non-negative* random variable X ,

$$\Pr\{X > tE[X]\} \leq 1/t .$$

Bernoulli and Binomial Random Variables. A Bernoulli(p) random variable is a random variable that is equal to 1 with probability p and 0 with probability $1 - p$. If X is a Bernoulli(p) random variable then $E[X] = p$. A binomial(p, n) random variable is the sum of n *independent* Bernoulli(p) random variables. If B is a Bernoulli(p, n) random variable then $E[B] = np$. Also, don't forget Chernoff's bounds:

$$\Pr\{B \geq (1 + \epsilon)np\} \leq e^{-\epsilon^2 np/3}$$

and

$$\Pr\{B \leq (1 - \epsilon)np\} \leq e^{-\epsilon^2 np/2}$$

Beware: Chernoff's bounds are only for binomial random variables. In particular you must make sure that the X_i s are independent!