

# CMPT 210: Probability and Computing

## Lecture 10

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## Back to throwing dice - Independent Events

**Q:** Suppose we throw two standard dice one after the other. What is the probability that we get two 6's in a row?

$E$  = We get a 6 in the second throw.  $F$  = We get a 6 in the first throw.  $E \cap F$  = we get two 6's in a row. We are computing  $\Pr[E \cap F]$ .  $\Pr[E] = \Pr[F] = \frac{1}{6}$ .

$$\Pr[E|F] = \frac{\Pr[E \cap F]}{\Pr[F]} \implies \Pr[E \cap F] = \Pr[E|F] \Pr[F].$$

Since the two dice are *independent*, knowing that we got a 6 in the first throw does not change the probability that we will get a 6 in the second throw. Hence,  $\Pr[E|F] = \Pr[E]$  (conditioning does not change the probability of the event).

$$\text{Hence, } \Pr[E \cap F] = \Pr[E|F] \Pr[F] = \Pr[E] \Pr[F] = \frac{1}{6} \frac{1}{6} = \frac{1}{36}.$$

# Independent Events

**Independent Events:** Events  $E$  and  $F$  are said to be independent, if knowledge that  $F$  has occurred does not change the probability that  $E$  occurs. Formally,

$$\Pr[E|F] = \Pr[E] \quad ; \quad \Pr[E \cap F] = \Pr[E] \Pr[F]$$

**Q:** I toss two independent, fair coins. What is the probability that I get the HT sequence?

Define  $E$  to be the event that I get a heads in the first toss, and  $F$  be the event that I get a tails in the second toss. Since the two coins are independent, events  $E$  and  $F$  are also independent.

$$\Pr[E \cap F] = \Pr[E] \Pr[F] = \frac{1}{2} \frac{1}{2} = \frac{1}{4}.$$

**Q:** I randomly choose a number from  $\{1, 2, \dots, 10\}$ .  $E$  is the event that the number I picked is a prime.  $F$  is the event that the number I picked is odd. Are  $E$  and  $F$  independent?

$\Pr[E] = \frac{2}{5}$ ,  $\Pr[F] = \frac{1}{2}$ ,  $\Pr[E \cap F] = \frac{3}{10}$ .  $\Pr[E \cap F] \neq \Pr[E] \Pr[F]$ . Another way:  $\Pr[E|F] = \frac{3}{5}$  and  $\Pr[E] = \frac{2}{5}$ , and hence  $\Pr[E|F] \neq \Pr[E]$ . Conditioning on  $F$  tell us that prime number cannot be 2, so it changes the probability of  $E$ .

## Independent Events - Example

**Q:** We have a machine that has 2 independent components. The machine breaks if *each* of its 2 components break. Suppose each component can break with probability  $p$ , what is the probability that the machine does not break?

Let  $E_1$  = Event that the first component breaks,  $E_2$  = Event that the second component breaks.  
 $M$  = Event that the machine breaks =  $E_1 \cap E_2$ .

$\Pr[M] = \Pr[E_1 \cap E_2]$ . Since the two components are independent,  $E_1$  and  $E_2$  are independent, meaning that  $\Pr[E_1 \cap E_2] = \Pr[E_1] \Pr[E_2] = p^2$ .

Probability that the machine does not break =  $\Pr[M^c] = 1 - \Pr[M] = 1 - p^2$ .

## Independent Events - Examples

**Q:** We have a new machine that has 2 independent components. The machine breaks if *either* of its 2 components break. Suppose each component can break with probability  $p$ , what is the probability that the machine breaks?

For this machine, let  $M'$  be the event that it breaks. In this case,  $\Pr[M'] = \Pr[E_1 \cup E_2]$ .

**Incorrect:** By the union rule for mutually exclusive events,  $\Pr[E_1 \cup E_2] = \Pr[E_1] + \Pr[E_2] = 2p$ .

**Mistake:** *Independence does not imply mutual exclusivity* and we can not use the union rule. Independence implies that for any two events  $E$  and  $F$ ,  $\Pr[E \cap F] = \Pr[E] \Pr[F]$ , while mutual exclusivity requires that  $\Pr[E \cap F] = 0$ .

Correct way 1:

$$\begin{aligned}\Pr[E_1 \cup E_2] &= \Pr[E_1] + \Pr[E_2] - \Pr[E_1 \cap E_2] && \text{(By the inclusion-exclusion rule)} \\ &= \Pr[E_1] + \Pr[E_2] - \Pr[E_1] \Pr[E_2] = 2p - p^2 && \text{(Since } E_1 \text{ and } E_2 \text{ are independent.)}\end{aligned}$$

## Independent Events - Examples

**Q:** We have a new machine that has 2 independent components. The machine breaks if *either* of its 2 components break. Suppose each component can break with probability  $p$ , what is the probability that the machine breaks?

Correct way 2:

$$\Pr[E_1 \cup E_2] = 1 - \Pr[(E_1 \cup E_2)^c] = 1 - \Pr[E_1^c \cap E_2^c]$$

(Complement of union of sets is equal to the intersection of the complements of sets)

$$= 1 - \Pr[E_1^c] \Pr[E_2^c] = 1 - (1 - p)^2 = 2p - p^2$$

(If  $E_1$  and  $E_2$  are independent, so are  $E_1^c$  and  $E_2^c$  (Proof on the next slide))

This implies that for the first machine, the probability of failure is  $p^2$  while for the second one, it is  $2p - p^2$ . Since  $p \leq 1$ ,  $p^2 \leq 2p - p^2$ , meaning that the first machine fails less often. This is intuitive since it fails only when *both* components fail.

## Independent Events - Examples

**Q:** Prove that if  $E_1$  and  $E_2$  are independent, so are  $E_1^c$  and  $E_2^c$ .

*Proof:*

$$\Pr[(E_1)^c \cap (E_2)^c] = \Pr[(E_1 \cup E_2)^c] = 1 - \Pr[E_1 \cup E_2] = 1 - \Pr[E_1] - \Pr[E_2] + \Pr[E_1 \cap E_2]$$

(By the inclusion-exclusion rule)

$$= 1 - \Pr[E_1] - \Pr[E_2] + \Pr[E_1] \Pr[E_2]$$

(Since  $E_1$  and  $E_2$  are independent)

$$\implies \Pr[(E_1)^c \cap (E_2)^c] = (1 - \Pr[E_1]) (1 - \Pr[E_2]) = \Pr[E_1^c] \Pr[E_2^c]$$

Hence, events  $E_1^c$  and  $E_2^c$  are independent.

Questions?



# Matrix Multiplication

Given two  $n \times n$  matrices –  $A$  and  $B$ , if  $C = AB$ , then,

$$C_{i,j} = \sum_{k=1}^n A_{i,k} B_{k,j}$$

Hence, in the worst case, computing  $C_{i,j}$  is an  $O(n)$  operation. There are  $n^2$  entries to fill in  $C$  and hence, in the absence of additional structure, matrix multiplication takes  $O(n^3)$  time.

There are non-trivial algorithms for doing matrix multiplication more efficiently:

- (Strassen, 1969) Requires  $O(n^{2.81})$  operations.
- (Coppersmith-Winograd, 1987) Requires  $O(n^{2.376})$  operations.
- (Alman-Williams, 2020) Requires  $O(n^{2.373})$  operations.
- Belief is that it can be done in time  $O(n^{2+\epsilon})$  for  $\epsilon > 0$ .

# Verifying Matrix Multiplication

As an example, let us focus on  $A, B$  being binary  $2 \times 2$  matrices.

Example:  $A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$  then  $C = AB = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$

**Objective:** Verify whether a matrix multiplication operation is correct.

**Trivial way:** Do the matrix multiplication ourselves, and verify it using  $O(n^3)$  (or  $O(n^{2.373})$ ) operations.

**Frievald's Algorithm:** Randomized algorithm to verify matrix multiplication with high probability in  $O(n^2)$  time.

## (Basic) Freivald's Algorithm

**Q:** For  $n \times n$  matrices  $A$ ,  $B$  and  $D$ , is  $D = AB$ ?

*Algorithm:*

1. Generate a random  $n$ -bit vector  $x$ , by making each bit  $x_i$  either 0 or 1 *independently* with probability  $\frac{1}{2}$ . E.g, for  $n = 2$ , toss a fair coin independently twice with the scheme – H is 0 and T is 1). If we get  $HT$ , then set  $x = [0; 1]$ .
2. Compute  $t = Bx$  and  $y = At = A(Bx)$  and  $z = Dx$ .
3. Output “yes” if  $y = z$  (all entries need to be equal), else output “no”.

**Computational complexity:** Step 1 can be done in  $O(n)$  time. Step 2 requires 3 matrix vector multiplications and can be done in  $O(n^2)$  time. Step 3 requires comparing two  $n$ -dimensional vectors and can be done in  $O(n)$  time. Hence, the total computational complexity is  $O(n^2)$ .

## (Basic) Frievald's Algorithm

Let us run the algorithm on an example. Suppose we have generated  $x = [1; 0]$

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad ; \quad B = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \quad ; \quad D = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$
$$Bx = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad ; \quad y = A(Bx) = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad ; \quad z = Dx = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Hence the algorithm will correctly output “no” since  $D \neq AB$ .

**Q:** Suppose we have generated  $x = [0; 0]$ . What is  $y$  and  $z$ ?

In this case,  $y = z$  and the algorithm will incorrectly output “yes” even though  $D \neq AB$ .

## (Basic) Frievald's Algorithm

Let us run the algorithm on an example. Suppose we have generated  $x = [1; 0]$ .

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad ; \quad B = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \quad ; \quad C = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$
$$Bx = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad ; \quad y = A(Bx) = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad ; \quad z = Cx = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Hence the algorithm will correctly output “yes” since  $C = AB$ .

**Q:** Suppose we have generated  $x = [0; 1]$ . What is  $y$  and  $z$ ?

In this case again,  $y = z$  and the algorithm will correctly output “yes”.

## (Basic) Freivald's Algorithm

Let us analyze the algorithm for general matrix multiplication.

**Case (i):** If  $D = AB$ , does the algorithm always output “yes”? Yes! Since  $D = AB$ , for any vector  $x$ ,  $Dx = ABx$ .

**Case (ii)** If  $D \neq AB$ , does the algorithm always output “no”?

**Claim:** For any input matrices  $A, B, D$  if  $D \neq AB$ , then the (Basic) Freivald's algorithm will output “no” with probability  $\geq \frac{1}{2}$ .

	Yes	No
$D = AB$	1	0
$D \neq AB$	$< \frac{1}{2}$	$\geq \frac{1}{2}$

**Table 1:** Probabilities for Basic Freivalds Algorithm

## (Basic) Frievald's Algorithm

*Proof:* If  $D \neq AB$ , we wish to compute the probability that algorithm outputs “yes” and prove that it less than  $\frac{1}{2}$ .

Define  $E := (AB - D)$  and  $r := Ex = (AB - D)x = y - z$ . If  $D \neq AB$ , then  $\exists(i, j)$  s.t.  $E_{i,j} \neq 0$ .

$$\begin{aligned}\Pr[\text{Algorithm outputs “yes”}] &= \Pr[y = z] = \Pr[r = \mathbf{0}] \\ &= \Pr[(r_1 = 0) \cap (r_2 = 0) \cap \dots \cap (r_i = 0) \cap \dots] \\ &= \Pr[(r_i = 0)] \Pr[(r_1 = 0) \cap (r_2 = 0) \cap \dots \cap (r_n = 0) | r_i = 0] \\ &\hspace{15em} (\text{By def. of conditional probability})\end{aligned}$$

$$\implies \Pr[\text{Algorithm outputs “yes”}] \leq \Pr[r_i = 0] \hspace{10em} (\text{Probabilities are in } [0, 1])$$

To complete the proof, on the next slide, we will prove that  $\Pr[r_i = 0] \leq \frac{1}{2}$ .

## (Basic) Freivald's Algorithm

$$r_i = \sum_{k=1}^n E_{i,k} x_k = E_{i,j} x_j + \sum_{k \neq j} E_{i,k} x_k = E_{i,j} x_j + \omega \quad (\omega := \sum_{k \neq j} E_{i,k} x_k)$$

$$\Pr[r_i = 0] = \Pr[r_i = 0 | \omega = 0] \Pr[\omega = 0] + \Pr[r_i = 0 | \omega \neq 0] \Pr[\omega \neq 0]$$

(By the law of total probability)

$$\Pr[r_i = 0 | \omega = 0] = \Pr[x_j = 0] = \frac{1}{2} \quad (\text{Since } E_{i,j} \neq 0 \text{ and } \Pr[x_j = 1] = \frac{1}{2})$$

$$\Pr[r_i = 0 | \omega \neq 0] = \Pr[(x_j = 1) \cap E_{i,j} = -\omega] = \Pr[(x_j = 1)] \Pr[E_{i,j} = -\omega | x_j = 1]$$

(By def. of conditional probability)

$$\implies \Pr[r_i = 0 | \omega \neq 0] \leq \Pr[(x_j = 1)] = \frac{1}{2} \quad (\text{Probabilities are in } [0, 1], \Pr[x_j = 1] = \frac{1}{2})$$

$$\implies \Pr[r_i = 0] \leq \frac{1}{2} \Pr[\omega = 0] + \frac{1}{2} \Pr[\omega \neq 0] = \frac{1}{2} \Pr[\omega = 0] + \frac{1}{2} [1 - \Pr[\omega = 0]] = \frac{1}{2}$$

( $\Pr[E^c] = 1 - \Pr[E]$ )

$$\implies \Pr[\text{Algorithm outputs "yes"}] \leq \Pr[r_i = 0] \leq \frac{1}{2}.$$



## (Basic) Freivald's Algorithm

Hence, if  $D \neq AB$ , the Algorithm outputs “yes” with probability  $\leq \frac{1}{2} \implies$  the Algorithm outputs “no” with probability  $\geq \frac{1}{2}$ .

In the worst case, the algorithm can be incorrect half the time! We promised the algorithm would return the correct answer with “high” probability close to 1.

A common trick in randomized algorithms is to have  $m$  independent trials of an algorithm and aggregate the answer in some way, reducing the probability of error, thus *amplifying the probability of success*.

Questions?

# Frievald's Algorithm

By repeating the *Basic Frievald's Algorithm*  $m$  times, we will amplify the probability of success. The resulting complete Frievald's Algorithm is given by:

- 1 Run the Basic Frievald's Algorithm for  $m$  independent runs.
- 2 If *any* run of the Basic Frievald's Algorithm outputs "no", output "no".
- 3 If *all* runs of the Basic Frievald's Algorithm output "yes", output "yes".

	Yes	No
$D = AB$	1	0
$D \neq AB$	$< \frac{1}{2^m}$	$\geq 1 - \frac{1}{2^m}$

Table 2: Probabilities for Frievald's Algorithm

If  $m = 20$ , then Frievald's algorithm will make mistake with probability  $1/2^{20} \approx 10^{-6}$ .

**Computational Complexity:**  $O(mn^2)$

# Probability Amplification

Consider a randomized algorithm  $\mathcal{A}$  that is supposed to solve a binary decision problem i.e. it is supposed to answer either Yes or No. It has a one-sided error – (i) if the true answer is Yes, then the algorithm  $\mathcal{A}$  correctly outputs Yes with probability 1, but (ii) if the true answer is No, the algorithm  $\mathcal{A}$  incorrectly outputs Yes with probability  $\leq \frac{1}{2}$ .

Let us define a new algorithm  $\mathcal{B}$  that runs algorithm  $\mathcal{A}$   $m$  times, and if *any* run of  $\mathcal{A}$  outputs No, algorithm  $\mathcal{B}$  outputs No. If *all* runs of  $\mathcal{A}$  output Yes, algorithm  $\mathcal{B}$  outputs Yes.

**Q:** What is the probability that algorithm  $\mathcal{B}$  correctly outputs Yes if the true answer is Yes, and correctly outputs No if the true answer is No?

## Probability Amplification - Analysis

$$\begin{aligned} & \Pr[\mathcal{B} \text{ outputs Yes} \mid \text{true answer is Yes}] \\ &= \Pr[\mathcal{A}_1 \text{ outputs Yes} \cap \mathcal{A}_2 \text{ outputs Yes} \cap \dots \cap \mathcal{A}_m \text{ outputs Yes} \mid \text{true answer is Yes}] \\ &= \prod_{i=1}^m \Pr[\mathcal{A}_i \text{ outputs Yes} \mid \text{true answer is Yes}] = 1 \end{aligned} \quad \text{(Independence of runs)}$$

$$\begin{aligned} & \Pr[\mathcal{B} \text{ outputs No} \mid \text{true answer is No}] \\ &= 1 - \Pr[\mathcal{B} \text{ outputs Yes} \mid \text{true answer is No}] \\ &= 1 - \Pr[\mathcal{A}_1 \text{ outputs Yes} \cap \mathcal{A}_2 \text{ outputs Yes} \cap \dots \cap \mathcal{A}_m \text{ outputs Yes} \mid \text{true answer is No}] \\ &= 1 - \prod_{i=1}^m \Pr[\mathcal{A}_i \text{ outputs Yes} \mid \text{true answer is No}] \geq 1 - \frac{1}{2^m}. \end{aligned}$$

When the true answer is Yes, both  $\mathcal{B}$  and  $\mathcal{A}$  correctly output Yes. When the true answer is No,  $\mathcal{A}$  incorrectly outputs Yes with probability  $< \frac{1}{2}$ , but  $\mathcal{B}$  incorrectly outputs Yes with probability  $< \frac{1}{2^m} \ll \frac{1}{2}$ . By repeating the experiment, we have “amplified” the probability of success.

Questions?