CMPT 210: Probability and Computing

Lecture 10

Sharan Vaswani February 3, 2023 ${f Q}$: 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).

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,

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\Pr[E|F] = \Pr[E]; \Pr[E \cap F] = \Pr[E] \Pr[F]
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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 $\{1, 2, ..., 10\}$. *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.

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:

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

Questions?

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 $\epsilon > 0$.

As an example, let us focus on A, B being binary 2×2 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.

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

Algorithm:

- Generate a random *n*-bit vector *x*, by making each bit *x_i* either 0 or 1 *independently* with probability 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 $O(n^2)$.

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? Ans: y = [0; 0] and z = [0; 0]. In this case, y = z and the algorithm will incorrectly output "yes" even though $D \neq AB$. 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? Ans: y = [1; 0] and z = [1; 0]. In this case again, y = z and the algorithm will correctly output "yes".

(Basic) Frievald'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) Frievald's algorithm will output "no" with probability $\geq \frac{1}{2}$.

YesNo
$$D = AB$$
10 $D \neq AB$ $< \frac{1}{2}$ $\geq \frac{1}{2}$

Table 1: Probabilities for Basic Frievalds 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. $E_{i,j} \neq 0$.

$$Pr[Algorithm outputs "yes"] = Pr[y = z] = Pr[r = 0]$$

=
$$Pr[(r_1 = 0) \cap (r_2 = 0) \cap \ldots \cap (r_i = 0) \cap \ldots]$$

=
$$Pr[(r_i = 0)] Pr[(r_1 = 0) \cap (r_2 = 0) \cap \ldots \cap (r_n = 0) | r_i = 0]$$

(By def. of conditional probability)

 \implies Pr[Algorithm outputs "yes"] \leq Pr[$r_i = 0$] (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) Frievald's Algorithm

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

12

 $\Pr[r_i = 0] = \Pr[r_i = 0 | \omega = 0] \Pr[\omega = 0] + \Pr[r_i = 0 | \omega \neq 0] \Pr[\omega \neq 0]$ (By the law of total probability)

$$\Pr[r_{i} = 0 | \omega = 0] = \Pr[x_{j} = 0] = \frac{1}{2}$$
(Since $E_{i,j} \neq 0$ 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)

$$\implies \Pr[r_i = 0 | \omega \neq 0] \leq \Pr[(x_j = 1)] = \frac{1}{2} \qquad (\text{Probabilities are in } [0, 1], \ \Pr[x_j = 1] = \frac{1}{2})$$
$$\implies \Pr[r_i = 0] \leq \frac{1}{2} \Pr[\omega = 0] + \frac{1}{2} \Pr[\omega \neq 0] = \frac{1}{2} \Pr[\omega = 0] + \frac{1}{2} [1 - \Pr[\omega = 0]] = \frac{1}{2} (\Pr[E^c] = 1 - \Pr[E])$$
$$\implies \Pr[\text{Algorithm outputs "yes"}] \leq \Pr[r_i = 0] \leq \frac{1}{2}.$$

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?