

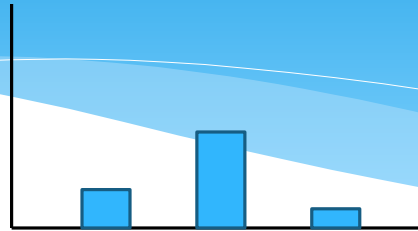
Review: Random Variables

- * A **random variable** represents a random quantity that depends on the outcome of a random experiment.
- * **Example:** Let **N** be the number of satisfied clauses of a 3CNF formula ϕ when we generate an assignment of ϕ by assigning its variables T/F independently and uniformly at random.

$$\phi = (x \vee x \vee x) \wedge (y \vee y \vee y) \wedge (\neg x \vee \neg y \vee \neg z)$$

Outcome	FFF	FFT	FTF	FTT	TFF	TFT	TTF	TTT
Sat?	NNY	NNY	NYN	NYN	YNY	YNY	YYY	YYN
N	1	1	2	2	2	2	3	2

Review: Distribution of an RV



- * The **probability** that a random variable is *equal* to a fixed value is the sum of the probabilities of all outcomes that result in that value occurring.
- * **Example:** Let N be the number of satisfied clauses of a 3CNF formula ϕ when we generate an assignment of ϕ by assigning its variables T/F independently and uniformly at random.

$$\phi = (x \vee x \vee x) \wedge (y \vee y \vee y) \wedge (\neg x \vee \neg y \vee \neg z)$$

Outcome	FFF	FFT	FTF	FTT	TFF	TFT	TTF	TTT
Sat?	NNY	NNY	NYN	NYN	YNY	YNY	YYY	YYN
N	1	1	2	2	2	2	3	2

$$\Pr[N = 1] = \frac{2}{8}$$

$$\Pr[N = 2] = \frac{5}{8}$$

$$\Pr[N = 3] = \frac{1}{8}$$

$$(\Pr[\text{each outcome}] = 1/8)$$

Review: Expected value of an RV

- * The **expected value** of a random variable is the weighted average of its values (weight of value = its probability).
- * **Example:** Let N be the number of satisfied clauses of a 3CNF formula ϕ when we generate an assignment of ϕ by assigning its variables T/F independently and uniformly at random.

$$\phi = (x \vee x \vee x) \wedge (y \vee y \vee y) \wedge (\neg x \vee \neg y \vee \neg z)$$

Outcome	FFF	FFT	FTF	FTT	TFF	TFT	TTF	TTT
Sat?	NNY	NNY	NYY	NYY	YNY	YNY	YYY	YYN
N	1	1	2	2	2	2	3	2

$$\Pr[N = 1] = \frac{2}{8} \quad \Pr[N = 2] = \frac{5}{8} \quad \Pr[N = 3] = \frac{1}{8}$$

$$\mathbb{E}[N] = 1 \cdot \frac{2}{8} + 2 \cdot \frac{5}{8} + 3 \cdot \frac{1}{8} = \frac{1}{8} (1 + 1 + 2 + 2 + 2 + 2 + 3 + 2)$$

On average, a random assignment satisfies $15/8$ clauses of ϕ .

Review: Linearity of Expectation

Observation: The number of clauses satisfied by the resulting assignment is $N = N_1 + N_2 + N_3 + \dots + N_m$.

- * **Linearity of expectation:** If $N = N_1 + N_2 + \dots + N_m$, then $\mathbb{E}[N] = \mathbb{E}[N_1] + \mathbb{E}[N_2] + \dots + \mathbb{E}[N_m]$
- * **Example:** Fix a 3CNF formula $\phi = C_1 \wedge C_2 \wedge \dots \wedge C_m$. Suppose we generate an assignment of ϕ by setting its variables to T/F independently, unif. at random.
 - * For $1 \leq i \leq m$, let N_i be the indicator random variable for whether clause C_i is satisfied by the assignment.

$$\phi = (x \vee x \vee x) \wedge (y \vee y \vee y) \wedge (\neg x \vee \neg y \vee \neg z)$$

Outcome	FFF	FFT	FTF	FTT	TFF	TFT	TTF	TTT
(N_1, N_2, N_3)	(0,0,1)	(0,0,1)	(0,1,1)	(0,1,1)	(1,0,1)	(1,0,1)	(1,1,1)	(1,1,0)
N	1	1	2	2	2	2	3	2

$$\mathbb{E}[N_1] = 1/2$$

$$\mathbb{E}[N_2] = 1/2$$

$$\mathbb{E}[N_3] = 7/8$$

$$\mathbb{E}[N] = \mathbb{E}[N_1] + \mathbb{E}[N_2] + \mathbb{E}[N_3] = 15/8$$

Review: Independence

- * An **event** is a set of outcomes.
- * **Example:** Let V be an RV for the value when rolling two dice.
 - * $V = 5$ is an event: the set of outcomes $\{(1,4), (2,3), (3,2), (4,1)\}$
 - * $\Pr[V = 5] = 4/36 = 1/9$
- * **Informally:** Two events are **independent** if the occurrence of one does not affect the probability of the other occurring.
- * **Formally:** Events A and B are **independent** if $\Pr[A \cap B] = \Pr[A] \cdot \Pr[B]$.
- * **Alternatively:** $\Pr[A|B] := \Pr[A \cap B]/\Pr[B]$ = probability of A given (conditioned on) B .
 - * A and B are **independent** if $\Pr[A \cap B] = \Pr[A]$.
- * A collection of random variables Z_1, \dots, Z_n is **independent** if for every $b_1, \dots, b_n : \Pr[Z_1 = b_1, \dots, Z_n = b_n] = \prod_i \Pr[Z_i = b_i]$.

Markov's Inequality

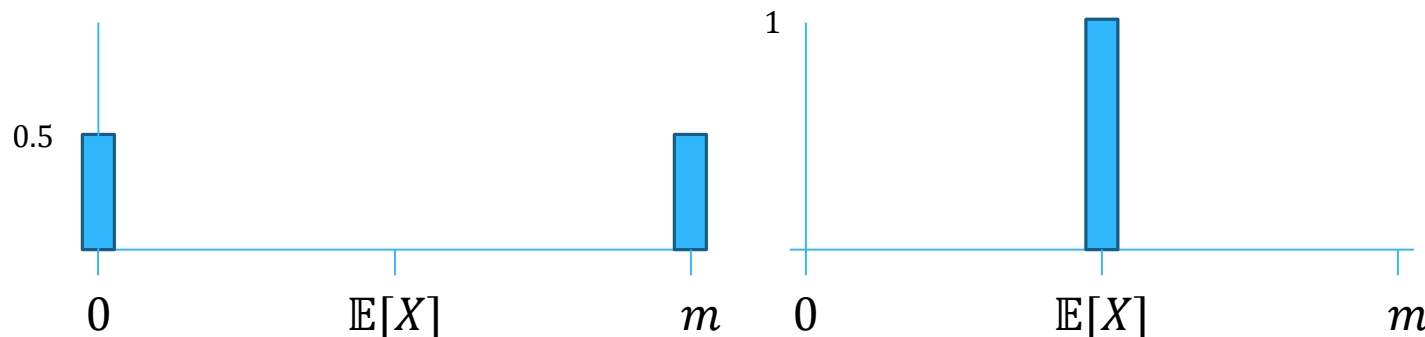
- * **Example:** The average score on the midterm was 60. What's the *maximum* fraction of students that got a score of at least 90 (assuming no negative scores)?
 - * 1/2? 6/9? 3/4? 99/100?
- * **Markov's Inequality:** If X is a non-negative random variable and $a > 0$, then $\Pr[X \geq a] \leq \mathbb{E}[X]/a$.

Variance

- * The **variance** of a random variable X is the average squared-distance of X from its mean, i.e.,

$$\mathbf{Var}(X) = \mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2] - \mathbb{E}[X]^2$$

- * The **standard deviation** is $\mathbf{SD}(X) = \sqrt{\mathbf{Var}(X)}$ (it's an upper bound on the average distance of X from $\mathbb{E}[X]$).



Chebyshev's Inequality

$$\mathbf{Var}(X) = \mathbb{E}[X^2] - \mathbb{E}[X]^2$$

- * **(Recall) Markov's Inequality:** For a non-negative RV X and $a > 0$:

$$\Pr[X \geq a] \leq \mathbb{E}[X]/a$$

- * **Chebyshev's Inequality:** For any RV X and $a, b > 0$:

$$\Pr[|X - \mathbb{E}[X]| \geq a] \leq \mathbf{Var}(X)/a^2$$

$$\Pr[|X - \mathbb{E}[X]| \geq b \cdot \mathbf{SD}(X)] \leq 1/b^2$$

(proof by applying Markov to $Y = (X - \mathbb{E}[X])^2$)

Chernoff-Hoeffding Bounds

(Often tighter than Chebyshev)

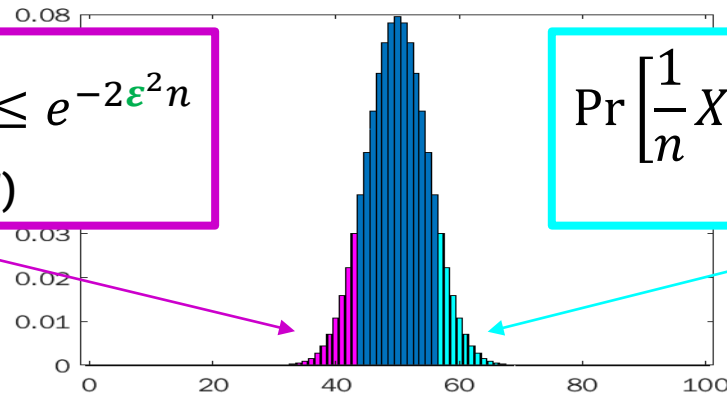
- * If $X = X_1 + X_2 + \dots + X_n$ is the sum of n *i.i.d.* RVs with each $X_i \in [0, 1]$, then, for any $\epsilon > 0$:

$$\Pr \left[\frac{1}{n} X \leq \mathbb{E}[X_i] - \epsilon \right] \leq e^{-2\epsilon^2 n}$$

(lower tail bound)

$$\Pr \left[\frac{1}{n} X \geq \mathbb{E}[X_i] + \epsilon \right] \leq e^{-2\epsilon^2 n}$$

(upper tail bound)



Note: These bounds work for *all* $n \geq 1$!