



Probability and Stochastic Processes

Lecture 23: Expectations of Simple, Non-Negative, and Arbitrary
Random Variables, Properties of Expectations

Karthik P. N.

Assistant Professor, Department of AI

Email: pnkarthik@ai.iith.ac.in

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Expectation of a Simple Random Variable

Definition (Expectation of a Simple RV)

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$. Let $X : \Omega \rightarrow \mathbb{R}$ be a random variable. Suppose that X is **simple** with the canonical representation

$$X = \sum_{i=1}^n a_i \mathbf{1}_{A_i}, \quad a_1, \dots, a_n \geq 0 \text{ distinct}, \quad A_1, \dots, A_n \in \mathcal{F} \text{ disjoint}.$$

Then, the **expectation of X** , denoted $\mathbb{E}[X]$, is defined as

$$\mathbb{E}[X] = \int_{\Omega} X \, d\mathbb{P} := \sum_{i=1}^n a_i \mathbb{P}(A_i).$$

- For any $A \in \mathcal{F}$,

$$\mathbb{E}[\mathbf{1}_A] = \mathbb{P}(A).$$

- The expectation of a simple RV is a non-negative and finite real number

Expectations of Non-Negative Random Variables

- Suppose X is **any non-negative random variable** (not necessarily simple)
- For each $n \in \mathbb{N}$, define X_n as

$$X_n = \sum_{\ell=0}^{n 2^n - 1} \frac{\ell}{2^n} \mathbf{1}_{\left\{ \frac{\ell}{2^n} \leq X < \frac{\ell+1}{2^n} \right\}} + n \mathbf{1}_{\{X \geq n\}},$$

$$X_n(\omega) = \begin{cases} 0, & 0 \leq X(\omega) < \frac{1}{2^n}, \\ \frac{1}{2^n}, & \frac{1}{2^n} \leq X(\omega) < \frac{2}{2^n}, \\ \vdots & \\ \frac{n 2^n - 1}{2^n}, & \frac{n 2^n - 1}{2^n} \leq X(\omega) < n, \\ n, & X(\omega) \geq n. \end{cases}.$$

- X_n is a **simple** random variable for each $n \in \mathbb{N}$ $\mathbb{E}[X_n]$ is well-defined for each $n \in \mathbb{N}$
- X_n can be represented compactly as

$$X_n = \frac{\lfloor 2^n X \rfloor}{2^n} \mathbf{1}_{\{X < n\}} + n \mathbf{1}_{\{X \geq n\}}$$

$$\mathbb{E}[X] = \int_{\Omega} X d\mathbb{P} := \lim_{n \rightarrow \infty} \mathbb{E}[X_n].$$

Properties of $\{X_n\}$

- X_n is a simple RV for each $n \in \mathbb{N}$
- **Monotonicity:** For each $\omega \in \Omega$, $X_n(\omega) \leq X_{n+1}(\omega) \quad \forall n \in \mathbb{N}$.
- **Pointwise convergence:** For each $\omega \in \Omega$, $\lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)$.
- **Monotonicity of expectations:** $\mathbb{E}[X_n] \leq \mathbb{E}[X_{n+1}]$ for all $n \in \mathbb{N}$ (exercise)
Therefore, $\lim_{n \rightarrow \infty} \mathbb{E}[X_n]$ exists
- $\lim_{n \rightarrow \infty} \mathbb{E}[X_n]$ may be equal to $+\infty$

Remark

If X is a non-negative random variable, then $0 \leq \mathbb{E}[X] \leq +\infty$.

Expectation of an Arbitrary Random Variable

- Suppose that X is **any arbitrary random variable** (potentially taking both +ve and -ve values)
- Define X_+ and X_- as

$$\forall \omega \in \Omega, \quad X_+(\omega) = \max\{X(\omega), 0\}, \quad X_-(\omega) = -\min\{X(\omega), 0\} = \max\{-X(\omega), 0\}.$$

- X_+ represents the **non-negative part** of X
 X_- represents the **negative part** of X
- **Exercise:** X_+ and X_- are random variables. Furthermore, they are **non-negative** RVs.
- $0 \leq \mathbb{E}[X_+] \leq +\infty, \quad 0 \leq \mathbb{E}[X_-] \leq +\infty$

Suppose X is an arbitrary random variable. Then, the **expectation** of X is defined as

$$\mathbb{E}[X] = \int_{\Omega} X \, d\mathbb{P} \quad := \quad \mathbb{E}[X_+] - \mathbb{E}[X_-],$$

provided both $\mathbb{E}[X_+]$ and $\mathbb{E}[X_-]$ are not $+\infty$ simultaneously.

If both $\mathbb{E}[X_+]$ and $\mathbb{E}[X_-]$ are equal to $+\infty$, then the expectation of X is **not defined**.

Properties of Expectations

Property 1

For any $A \in \mathcal{F}$ such that $\mathbb{P}(A) = 0$, $\int_A X \, d\mathbb{P} = \mathbb{E}[X \cdot \mathbf{1}_A] = 0$.

Proof:

- Suppose X is **simple** with the canonical representation

$$X = \sum_{i=1}^n a_i \mathbf{1}_{A_i}, \quad a_1, \dots, a_n \text{ distinct}, \quad A_1, \dots, A_n \text{ disjoint.}$$

- Then, $X \mathbf{1}_A$ is also a simple random variable, with the canonical representation

$$X \mathbf{1}_A = \sum_{i=1}^n a_i \mathbf{1}_{A_i} \mathbf{1}_A = \sum_{i=1}^n a_i \mathbf{1}_{A_i \cap A}.$$

- Taking expectations, we get

$$\mathbb{E}[X \mathbf{1}_A] = \sum_{i=1}^n a_i \mathbb{P}(A_i \cap A) = 0 \quad (\mathbb{P}(A_i) = 0)$$

Property 1

For any $A \in \mathcal{F}$ such that $\mathbb{P}(A) = 0$, $\int_A X \, d\mathbb{P} = \mathbb{E}[X \cdot \mathbf{1}_A] = 0$.

Proof:

- Suppose X is **non-negative**
- For each $n \in \mathbb{N}$, define X_n as

$$X_n = \sum_{\ell=0}^{n 2^n - 1} \frac{\ell}{2^n} \mathbf{1}_{\left\{\frac{\ell}{2^n} \leq X < \frac{\ell+1}{2^n}\right\}} + n \mathbf{1}_{\{X \geq n\}}, \quad X_n \mathbf{1}_A = \left(\sum_{\ell=0}^{n 2^n - 1} \frac{\ell}{2^n} \mathbf{1}_{\left\{\frac{\ell}{2^n} \leq X < \frac{\ell+1}{2^n}\right\}} + n \mathbf{1}_{\{X \geq n\}} \right) \mathbf{1}_A.$$

- We have

$$\mathbb{E}[X \mathbf{1}_A] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n \mathbf{1}_A] = \lim_{n \rightarrow \infty} 0 = 0.$$

Property 1

For any $A \in \mathcal{F}$ such that $\mathbb{P}(A) = 0$, $\int_A X \, d\mathbb{P} = \mathbb{E}[X \cdot \mathbf{1}_A] = 0$.

Proof:

- Suppose X is **arbitrary**
- Define X_+ and X_- as

$$X_+ = \max\{X, 0\}, \quad X_- = -\min\{X, 0\}$$

- Observe that

$$(X \mathbf{1}_A)_+ = X_+ \mathbf{1}_A, \quad (X \mathbf{1}_A)_- = X_- \mathbf{1}_A$$

- We therefore have

$$\mathbb{E}[X \mathbf{1}_A] = \mathbb{E}[(X \mathbf{1}_A)_+] - \mathbb{E}[(X \mathbf{1}_A)_-] = \mathbb{E}[X_+ \mathbf{1}_A] - \mathbb{E}[X_- \mathbf{1}_A] = 0 - 0 = 0.$$

Property 2

For any $A \in \mathcal{F}$ such that $\mathbb{P}(A) = 1$, $\int_A X \, d\mathbb{P} = \mathbb{E}[X \cdot \mathbf{1}_A] = \mathbb{E}[X]$.

Proof:

- Suppose X is **simple** with the canonical representation

$$X = \sum_{i=1}^n a_i \mathbf{1}_{A_i}, \quad a_1, \dots, a_n \text{ distinct}, \quad A_1, \dots, A_n \text{ disjoint}.$$

- Then, $X \mathbf{1}_A$ is also a simple random variable, with the canonical representation

$$X \mathbf{1}_A = \sum_{i=1}^n a_i \mathbf{1}_{A_i} \mathbf{1}_A = \sum_{i=1}^n a_i \mathbf{1}_{A_i \cap A}.$$

- Taking expectations, we get

$$\mathbb{E}[X \mathbf{1}_A] = \sum_{i=1}^n a_i \mathbb{P}(A_i \cap A) = \sum_{i=1}^n a_i \mathbb{P}(A_i) = \mathbb{E}[X].$$

Property 2

For any $A \in \mathcal{F}$ such that $\mathbb{P}(A) = 1$, $\int_A X \, d\mathbb{P} = \mathbb{E}[X \cdot \mathbf{1}_A] = \mathbb{E}[X]$.

Proof:

- Suppose X is **non-negative**
- For each $n \in \mathbb{N}$, define X_n as

$$X_n = \sum_{\ell=0}^{n 2^n - 1} \frac{\ell}{2^n} \mathbf{1}_{\left\{\frac{\ell}{2^n} \leq X < \frac{\ell+1}{2^n}\right\}} + n \mathbf{1}_{\{X \geq n\}}, \quad X_n \mathbf{1}_A = \left(\sum_{\ell=0}^{n 2^n - 1} \frac{\ell}{2^n} \mathbf{1}_{\left\{\frac{\ell}{2^n} \leq X < \frac{\ell+1}{2^n}\right\}} + n \mathbf{1}_{\{X \geq n\}} \right) \mathbf{1}_A.$$

- We have

$$\mathbb{E}[X \mathbf{1}_A] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n \mathbf{1}_A] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n] = \mathbb{E}[X].$$

Property 2

For any $A \in \mathcal{F}$ such that $\mathbb{P}(A) = 0$, $\int_A X \, d\mathbb{P} = \mathbb{E}[X \cdot \mathbf{1}_A] = 0$.

Proof:

- Suppose X is **arbitrary**
- Define X_+ and X_- as

$$X_+ = \max\{X, 0\}, \quad X_- = -\min\{X, 0\}$$

- Observe that

$$(X \mathbf{1}_A)_+ = X_+ \mathbf{1}_A, \quad (X \mathbf{1}_A)_- = X_- \mathbf{1}_A$$

- We therefore have

$$\mathbb{E}[X \mathbf{1}_A] = \mathbb{E}[(X \mathbf{1}_A)_+] - \mathbb{E}[(X \mathbf{1}_A)_-] = \mathbb{E}[X_+ \mathbf{1}_A] - \mathbb{E}[X_- \mathbf{1}_A] = \mathbb{E}[X_+] - \mathbb{E}[X_-] = \mathbb{E}[X].$$

Property 3

If $\mathbb{P}(\{X \geq Y\}) = 1$, then

$$\mathbb{E}[X] \geq \mathbb{E}[Y].$$

Proof:

- Define $A = \{X \geq Y\}$. It is given that $\mathbb{P}(A) = 1$.
- We have

$$\mathbb{E}[X] = \mathbb{E}[X \mathbf{1}_A] = \int_A X(\omega) \, d\mathbb{P}(\omega) \geq \int_A Y(\omega) \, d\mathbb{P}(\omega) = \mathbb{E}[Y \mathbf{1}_A] = \mathbb{E}[Y].$$

Property 4

If $\mathbb{P}(\{X = Y\}) = 1$, then

$$\mathbb{E}[X] = \mathbb{E}[Y].$$

Proof:

- Define $A = \{X = Y\}$. It is given that $\mathbb{P}(A) = 1$.
- We have

$$\mathbb{E}[X] = \mathbb{E}[X \mathbf{1}_A] = \int_A X(\omega) \, d\mathbb{P}(\omega) = \int_A Y(\omega) \, d\mathbb{P}(\omega) = \mathbb{E}[Y \mathbf{1}_A] = \mathbb{E}[Y].$$

Property 5

If X is a **non-negative RV**, then

$$\mathbb{E}[X] = 0 \quad \Longleftrightarrow \quad \mathbb{P}(\{X = 0\}) = 1.$$

Proof: (\Leftarrow)

- Let $A = \{X = 0\}$, and suppose that $\mathbb{P}(A) = 1$
- Then,

$$\mathbb{E}[X] = \mathbb{E}[X \mathbf{1}_A] = \int_{\{X=0\}} X(\omega) \, d\mathbb{P}(\omega) = \int_{\{X=0\}} 0 \, d\mathbb{P}(\omega) = 0.$$

Property 5

If X is a **non-negative RV**, then

$$\mathbb{E}[X] = 0 \quad \Longleftrightarrow \quad \mathbb{P}(\{X = 0\}) = 1.$$

Proof: (\implies)

- Suppose X is a **simple** RV with canonical representation

$$X = \sum_{i=1}^m a_i \mathbf{1}_{A_i}, \quad a_1, \dots, a_m \text{ distinct}, \quad A_1, \dots, A_m \text{ disjoint and partition of } \Omega.$$

Furthermore, suppose that $\mathbb{E}[X] = 0$

- We must have $\sum_{i=1}^m \mathbb{P}(A_i) = 1$
- Then, we have

$$0 = \mathbb{E}[X] = \sum_{i=1}^m a_i \mathbb{P}(A_i) \quad \implies \quad a_i \mathbb{P}(A_i) = 0 \quad \forall i \in \{1, \dots, m\}.$$

- There exists a unique $i^* \in \{1, \dots, m\}$ such that $a_{i^*} = 0$ and $A_{i^*} = \{X = a_{i^*}\}$ has probability 1

Property 5

If X is a **non-negative RV**, then

$$\mathbb{E}[X] = 0 \iff \mathbb{P}(\{X = 0\}) = 1.$$

Proof: (\implies)

- Suppose X is a **non-negative** RV with $\mathbb{E}[X] = 0$
- Let $\{X_n\}$ be the associated quantization sequence, with

$$0 \leq \mathbb{E}[X_1] \leq \mathbb{E}[X_2] \leq \dots \quad \lim_{n \rightarrow \infty} \mathbb{E}[X_n] = \mathbb{E}[X] = \mathbf{0}$$

- This implies that

$$\mathbb{E}[X_n] = 0 \quad \forall n \in \mathbb{N}.$$

- For each $n \in \mathbb{N}$, the set $A_n = \{X_n = 0\}$ has probability 1, and hence $\mathbb{P}(\cap_{n \in \mathbb{N}} A_n) = 1$
- Observe that

$$\forall \omega \in \Omega, \quad \lim_{n \rightarrow \infty} X_n(\omega) = X(\omega), \quad \{X = 0\} \supseteq \bigcap_{n \in \mathbb{N}} \{X_n = 0\}, \quad \mathbb{P}(\{X = 0\}) \geq \mathbb{P}\left(\bigcap_{n \in \mathbb{N}} A_n\right) = 1.$$

Property 6 (Linearity of Expectations)

For any two random variables X, Y ,

$$\mathbb{E}[\alpha X] = \alpha \mathbb{E}[X] \quad \forall \alpha \in \mathbb{R}, \quad \mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y],$$

whenever the right-hand sides are well-defined

Proof:

- For any $\alpha \in \mathbb{R}$,

$$\mathbb{E}[\alpha X] = \int_{\Omega} \alpha X(\omega) \, d\mathbb{P}(\omega) = \alpha \int_{\Omega} X(\omega) \, d\mathbb{P}(\omega) = \alpha \mathbb{E}[X].$$

Property 6 (Linearity of Expectations)

For any two random variables X, Y ,

$$\mathbb{E}[\alpha X] = \alpha \mathbb{E}[X] \quad \forall \alpha \in \mathbb{R}, \quad \mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y],$$

whenever the right-hand sides are well-defined

Proof:

- Suppose X and Y are **simple** RVs with canonical representations

$$X = \sum_{i=1}^m a_i \mathbf{1}_{A_i}, \quad Y = \sum_{j=1}^n b_j \mathbf{1}_{B_j},$$

where $\{A_1, \dots, A_m\}$ and $\{B_1, \dots, B_n\}$ **each** partition Ω

- Then, $X + Y$ has the representation

$$X + Y = \sum_{i=1}^m \sum_{j=1}^n (a_i + b_j) \mathbf{1}_{A_i \cap B_j}$$

- We may combine similar $(a_i + b_j)$ terms to bring $X + Y$ to canonical form

Property 6 (Linearity of Expectations)

For any two random variables X, Y ,

$$\mathbb{E}[\alpha X] = \alpha \mathbb{E}[X] \quad \forall \alpha \in \mathbb{R}, \quad \mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y],$$

whenever the right-hand sides are well-defined

Proof:

- We then have

$$\begin{aligned} \mathbb{E}[X + Y] &= \sum_{i=1}^m \sum_{j=1}^n (a_i + b_j) \mathbb{P}(A_i \cap B_j) = \sum_{i=1}^m a_i \underbrace{\sum_{j=1}^n \mathbb{P}(A_i \cap B_j)}_{=\mathbb{P}(A_i)} + \sum_{j=1}^n b_j \underbrace{\sum_{i=1}^m \mathbb{P}(A_i \cap B_j)}_{=\mathbb{P}(B_j)} \\ &= \sum_{i=1}^m a_i \mathbb{P}(A_i) + \sum_{j=1}^n b_j \mathbb{P}(B_j) = \mathbb{E}[X] + \mathbb{E}[Y]. \end{aligned}$$

Property 6 (Linearity of Expectations)

For any two random variables X, Y ,

$$\mathbb{E}[\alpha X] = \alpha \mathbb{E}[X] \quad \forall \alpha \in \mathbb{R}, \quad \mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y],$$

whenever the right-hand sides are well-defined

Proof:

- Suppose X and Y are **non-negative** random variables
- Let $\{X_n\}$ and $\{Y_n\}$ be the associated quantization sequences, with

$$\forall \omega \in \Omega, \quad \lim_{n \rightarrow \infty} X_n(\omega) + Y_n(\omega) = X(\omega) + Y(\omega), \quad \lim_{n \rightarrow \infty} \mathbb{E}[X_n + Y_n] = \mathbb{E}[X + Y],$$

- Then,

$$\mathbb{E}[X + Y] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n + Y_n] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n] + \lim_{n \rightarrow \infty} \mathbb{E}[Y_n] = \mathbb{E}[X] + \mathbb{E}[Y].$$

To be continued...