



Design and Analysis
of Algorithms I

Probability Review

Part I

Topics Covered

- Sample spaces
- Events
- Random variables
- Expectation
- Linearity of Expectation

See also:

- Lehman-Leighton notes (free PDF)
- Wikibook on Discrete Probability

Concept #1 – Sample Spaces

Sample space Ω = "all possible outcomes"

[in algorithms, Ω is usually finite]

Also: each outcome $i \in \Omega$ has a probability $p(i) \geq 0$.

Constraint: $\sum_{i \in \Omega} p(i) = 1$.

Example #1: Rolling 2 dice. $\Omega = \{(1,1), (2,1), (3,1), \dots, (5,6), (6,6)\}$
and $p(i) = 1/36$ for all $i \in \Omega$.
36 ordered pairs

Example #2: choosing a random pivot in after QuickSort call.

$\Omega = \{1, 2, \dots, n\}$ (index of pivot) and $p(i) = 1/n$ for all $i \in \Omega$.

Concept #2 – Events

An event is a subset $S \subseteq \Omega$.

The probability of an event S is $\sum_{i \in S} p(i)$.

Consider the event (i.e., the subset of outcomes for which) “the sum of the two dice is 7”. What is the probability of this event?

☐ $1/36$

☐ $1/12$

☒ $1/6$

☐ $1/2$

$$S = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$$

$$\underline{P(S) = 6/36 = 1/6}$$

Consider the event (i.e., the subset of outcomes for which) “the chosen pivot gives a 25-75 split of better”. What is the probability of this event?

☐ $1/n$

☐ $1/4$

☒ $1/2$

☐ $3/4$

$$S = \left\{ \left(\frac{n}{4} + 1\right)^{\text{th}} \text{ smallest element}, \dots, \left(\frac{3n}{4}\right)^{\text{th}} \text{ smallest element} \right\}$$

$$P_r[S] = \frac{n/2}{n} = \frac{1}{2}$$

Concept #2 – Events

An event is a subset $S \subseteq \Omega$.

The probability of an event S is $\sum_{i \in S} p(i)$.

Ex #1: sum of dice = 7. $S = \{(1,1), (2,1), (3,1), \dots, (5,6), (6,6)\}$.
 $P(S) = 6/36 = 1/6$.

Ex #2: pivot gives 25-75 split of letters.
 $S = \{(\frac{n}{4} + 1)^{\text{th}} \text{ smallest element}, \dots, (\frac{3n}{4})^{\text{th}} \text{ smallest element}\}$.
 $P(S) = \frac{n/2}{n} = 1/2$.

Concept #3 - Random Variables

A random variable X is a real-valued function
 $X: \Omega \rightarrow \mathbb{R}$.

Ex #1: Sum of the two dice.

Ex #2: Size of subarray passed to 1st recursive call.

Concept #4 - Expectation

Let $X: \Omega \rightarrow \mathbb{R}$ be a random variable.

The expectation $E[X]$ of X = average value of X

$$= \sum_{i \in \Omega} X(i) \cdot p(i)$$

What is the expectation of the sum of two dice?

☐ 6.5

☒ 7

☐ 7.5

☐ 8

Which of the following is closest to the expectation of the size of the subarray passed to the first recursive call in QuickSort?

- ☐ $n/4$
- ☐ $n/3$
- ☒ $n/2$
- ☐ $3n/4$

let X = subarray size

$$\begin{aligned} \text{Then } E[X] &= \frac{1}{n} \cdot 0 + \frac{1}{n} \cdot 1 + \frac{1}{n} \cdot 2 + \dots + \frac{1}{n} \cdot (n-1) \\ &= \frac{(n-1)}{2} \end{aligned}$$

Concept #4 - Expectation

Let $X: \Omega \rightarrow \mathbb{R}$ be a random variable.

The expectation $E[X]$ of X = average value of X

$$= \sum_{i \in \Omega} X(i) \cdot p(i)$$

Ex#1: if X = sum of 2 dice, $E[X] = 7$.

Ex#2: if X = 1st subarray size in QuickSort,
 $E[X] = (n-1)/2$.

Concept #5 – Linearity of Expectation

Claim [LIN EXP] : let X_1, \dots, X_n be random variables defined on Ω . Then:

$$E\left[\sum_{j=1}^n X_j\right] = \sum_{j=1}^n E[X_j]$$

CRUCIALLY:
HOLDS EVEN WHEN
 X_j 's ARE NOT
INDEPENDENT!

Ex 1: if X_1, X_2 = the two dice, then
 $E[X_j] = 1/6 (1+2+3+4+5+6) = 3.5$

By LIN EXP: $E[X_1] + E[X_2] = E[X_1 + X_2] = 3.5 + 3.5 = 7.$

[WOULD FAIL IF
REPLACES SUMS
WITH PRODUCTS]

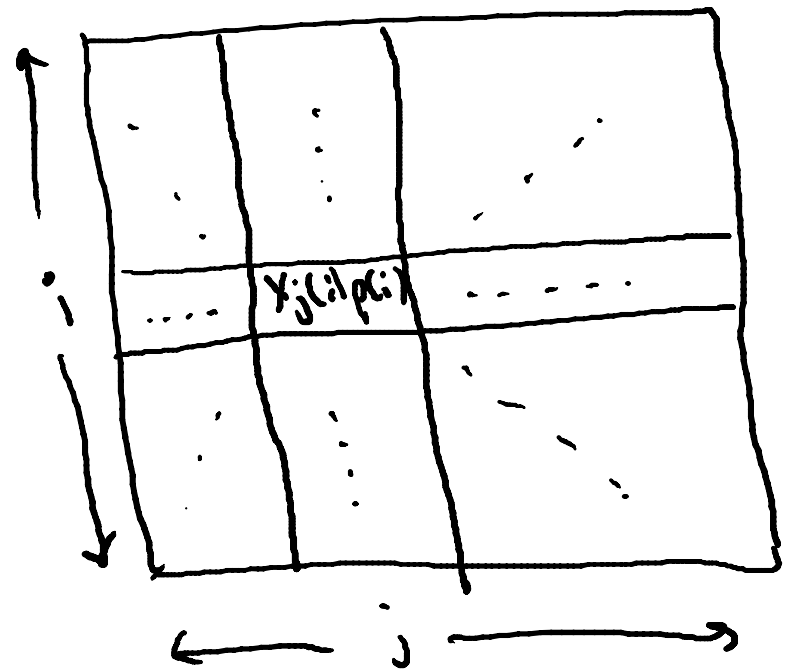
Linearity of Expectation (Proof)

$$\sum_{j=1}^n E[X_j] = \sum_{j=1}^n \sum_{i \in \Omega} \underline{X_j(i) p(i)}$$

$$\stackrel{(\circledast)}{=} \sum_{i \in \Omega} \sum_{j=1}^n \underline{X_j(i) p(i)}$$

$$= \sum_{i \in \Omega} p(i) \left(\sum_{j=1}^n X_j(i) \right)$$

$$= E \left[\sum_{j=1}^n X_j \right] \quad \text{QED!}$$



Example: Load Balancing

Problem: need to assign n processes to n servers.

Proposed Solution: assign each process to a random server.

Question: what is expected number of processes assigned to a server?

Load Balancing Solution

Sample space Ω = all n^m assignments of processes to servers, each equally likely.

Let Y = total number of processes assigned to the first server.

Goal: Compute $E[Y]$.

Let $X_j = \begin{cases} 1 & \text{if } j^{\text{th}} \text{ process assigned to first server} \\ 0 & \text{otherwise} \end{cases}$

↙ "indicator random variable"

Note: $Y = \sum_{j=1}^n X_j$

Load Balancing Solution (con'd)

We have

$$E[Y] = E\left[\sum_{j=1}^n X_j\right]$$

$$(\text{since } Y = \sum_{j=1}^n X_j)$$

$$= \sum_{j=1}^n E[X_j]$$

(LINEAR EXP)

$$= \sum_{j=1}^n (\cancel{\Pr[X_j=0]} \cdot 0 + \Pr[X_j=1] \cdot 1)$$

$= \frac{1}{n}$ (servers chosen uniformly at random)

$$= \sum_{j=1}^n \frac{1}{n}$$

$$= 1.$$