



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. $\Omega = \{(1,1), (2,1), (3,1), \dots, (5,6), (6,6)\}$

Example #2 : Choosing a random pivot in outer QuickSort call.

$\Omega = \{1,2,3,\dots,n\}$ (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?

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

$$\Pr[S] = 6/36 = 1/6$$

$1/36$

$1/12$

$1/6$

$1/2$

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?

$1/n$

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

$1/4$

$\Pr[S] = (n/2)/n = 1/2$

$1/2$

$3/4$

Concept #2 – Events

An event is a subset

The probability of an event S is

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

Ex#2 : pivot gives 25-75 split or better.
 $S = \{(n/4+1)^{\text{th}} \text{ smallest element}, \dots, (3n/4)^{\text{th}} \text{ smallest element}\}$
 $\Pr[S] = (n/2)/n = 1/2$

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?

Let X = subarray size

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

- $n/4$
- $n/3$
- $n/2$
- $3n/4$

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 : Sum of the two dice, $E[X] = 7$

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

$$E[X] = (n-1)/2$$

Concept #5 – Linearity of Expectation

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

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

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

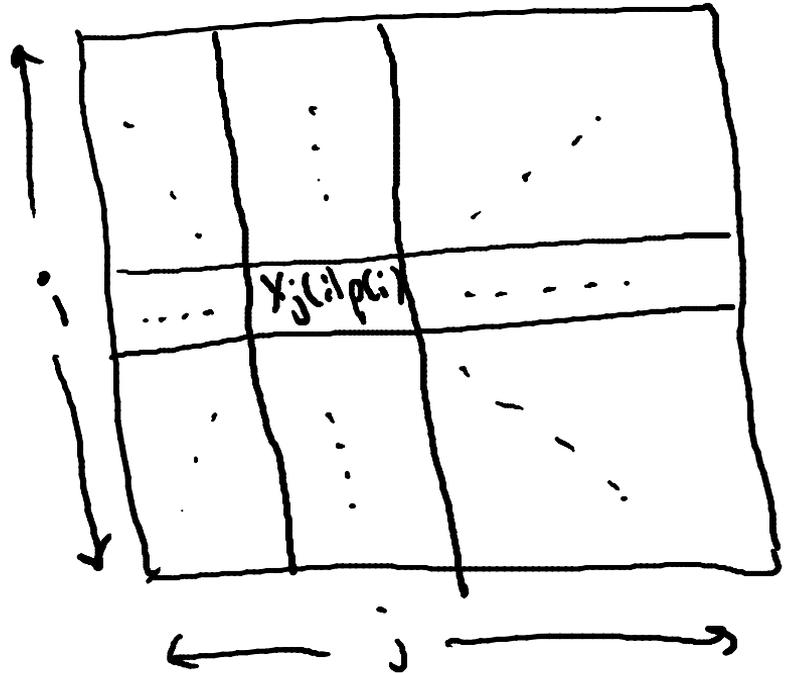
By LIN EXP : $E[X_1+X_2] = E[X_1] + E[X_2] = 3.5 + 3.5 = 7$

CRUCIALLY:
HOLDS EVEN WHEN
 X_j 's ARE NOT
INDEPENDENT!
[WOULD FAIL IF
REPLACE SUMS WITH
PRODUCTS]

Linearity of Expectation (Proof)

$$\begin{aligned}\sum_{j=1}^n E[X_j] &= \sum_{j=1}^n \sum_{i \in \Omega} X_j(i) p(i) \\ &= \sum_{i \in \Omega} \sum_{j=1}^n X_j(i) p(i) \\ &= \sum_{i \in \Omega} p(i) \sum_{j=1}^n X_j(i) \\ &= E\left[\sum_{j=1}^n X_j\right]\end{aligned}$$

Q.E.D.



Example: Load Balancing

Problem : need to assign n processes to n servers.

Proposed Solution : assign each process to a random server

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

Load Balancing Solution

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

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

Goal : compute $E[Y]$

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

“indicator random variable”

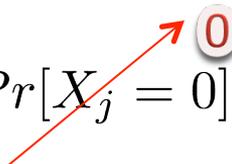


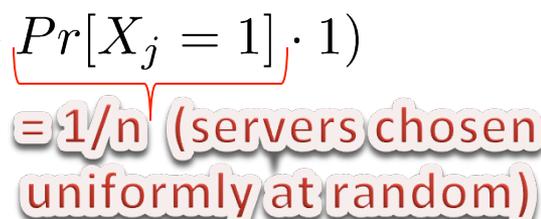
Note $Y = \sum_{j=1}^n X_j$

Load Balancing Solution (con'd)

We have

$$\begin{aligned} E[Y] &= E\left[\sum_{j=1}^n X_j\right] \\ &= \sum_{j=1}^n E[X_j] \\ &= \sum_{j=1}^n (Pr[X_j = 0] \cdot 0 + Pr[X_j = 1] \cdot 1) \\ &= \sum_{j=1}^n \frac{1}{n} = 1 \end{aligned}$$

 **0**

 **= 1/n (servers chosen uniformly at random)**