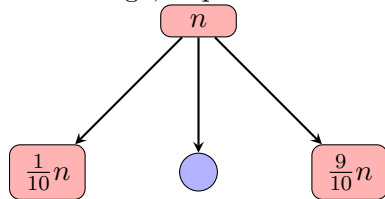


Algorithms

0.1 The Average Case

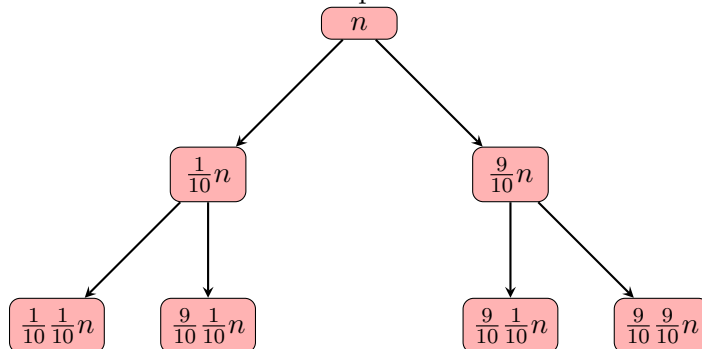
On average, a quick sort is close to its base case.



0.1.1 Recurrence

$$T(n) \leq T\left(\frac{9}{10}n\right) + T\left(\frac{1}{10}n\right) + C \cdot n$$

Recurrence Tree for a 9-1 Split:



The left will reach the bottom quickly, the right will take longer.

In general, any split that is a constant fraction will be in the same complexity category as its best case because it is not dependent on n .

The best split is when the pivot is in the middle.

The worst split is when nothing is on one side of the pivot.

0.1.2 Alternating Tree

Best Case Scenario:

$$\frac{n-1}{2} + \frac{n-1}{2} = n-1 \in O(n)$$

When the splits rotate from worst to best and back, it is the same time complexity as the best case..

0.1.3 Randomized Quicksort

Choose a random item from the list, rather than taking the first item as a point. Do this by grabbing a random item and swapping it with the first item.

0.2 Statistics Recap

The **Sample Space** S is the set of all possible outcomes.

$$S = \{S_1, S_2, S_3, \dots, S_k\}$$

An **Event** is a subset of a sample space, such as the event of “getting at least 1 heads” of the sample space “flipping two coins”.

Probability is the likelihood of an event.

$P(Event)$ = The sum of basic outcomes in the event.

1. Probability is always between 0 and 1
2. For any event $E \in S$, $P(E) = \sum_{S_i \in E} P(S_i)$
3. $P(S) = 1$, $P(\emptyset) = 0$

0.2.1 Discrete Random Variables

Every basic outcome is matched to a real number.

$$X : S \rightarrow \mathbb{R}$$

$$s \in S : X(s) = x$$

Probability Density Function:

$$f(x) = P(X = x)$$

$$0 \leq P(X = x) \leq 1$$

$$\sum_{i=1}^k P(X = x_i) = 1$$

Example:

Define X to be the highest result of 2 dice rolled.

$P(X = 5)$ would be the chance that the highest roll is 5.

The set defined by $X = 5$:

$$\begin{bmatrix} 15 & 51 & 25 \\ 52 & 35 & 53 \\ 45 & 54 & 55 \end{bmatrix}$$

$$P(X = 5) = \frac{1}{36} \cdot 9 = \frac{1}{4}$$

The **Expected Value of X** , or $E[X]$, is

$$E[X] = \sum_{i=1}^k X_i \cdot P(X_i)$$

Example:

In a game where you flip 2 coins, getting 3 points for heads and losing 2 points for tails, X represents the payout from a single round.

$$E[X] = 6 \cdot P(H, H) + 1 \cdot P(H, T) + 1 \cdot P(T, H) + -4 \cdot P(T, T) = 1$$

0.2.2 Indicator Random Variable

A way to go between P and E.

Example:

Flipping a coin n times, what is the expected number of heads?

Let C be the random variable representing the number of heads.

$C_i = 0$ if tails, $C_i = 1$ for i from 1 to n

$$C = \sum_{i=1}^n C_i$$

$$P(C_i = 1) = \frac{1}{2}, \quad P(C_i = 0) = \frac{1}{2}$$

$$E[C_i] = 0 \cdot P(C_i = 0) + 1 \cdot P(C_i = 1) = \frac{1}{2}$$

$$E[C] = E\left[\sum_{i=1}^n C_i\right] = \frac{n}{2}$$