



Data Structures and Algorithms

Probability in Computer Science

CS 225
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Randomization in Algorithms

1. Assume input data is random to estimate average-case performance
2. Use randomness inside algorithm to estimate expected running time
3. Use randomness inside algorithm to approximate solution in fixed time



Quicksort

```
Quicksort(A[1...n])
```

```
    if n ≤ 1
```

```
        return
```

```
    p = random(1-n)
```

```
    r = Partition(A[1...n], p)
```

```
    Quicksort(A[1...r-1])
```

```
    Quicksort(A[r+1...n])
```

```
Partition(A[1...n], p)
```

```
    swap(A[p], A[n])
```

```
    l = 0
```

```
    for i 1 to n-1
```

```
        if A[i] < A[n]
```

```
            l = l + 1
```

```
            swap(A[l], A[i])
```

```
    l = l + 1
```

```
    swap(A[n], A[l])
```

```
    return l
```

Expectation Analysis: Randomized Quicksort

6	1	0	3	7	9	2	4
---	---	---	---	---	---	---	---

1	0	3	2	4	9	6	7
---	---	---	---	---	---	---	---

1	0	3	2	4	9	6	7
---	---	---	---	---	---	---	---

1	0	2	3	4	6	7	9
---	---	---	---	---	---	---	---

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---	---	---	---	---	---	---	---

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0	1	2	3	4	6	7	9
---	---	---	---	---	---	---	---

0	1	2	3	4	5	6	7
---	---	---	---	---	---	---	---

0	1	2	3	4	5	6	7
---	---	---	---	---	---	---	---

0	1	2	3	4	5	6	7
---	---	---	---	---	---	---	---

...

0	1	2	3	4	5	6	7
---	---	---	---	---	---	---	---

Expectation Analysis: Randomized Quicksort

In **randomized quicksort**, the selection of the pivot is random.

Claim: The expected comparisons is $O(n \log n)$ *for any input!*

Let X be the total comparisons and X_{ij} be an **indicator variable**:

$$X_{ij} = \begin{cases} 1 & \text{if } i\text{th object compared to } j\text{th} \\ 0 & \text{if } i\text{th object not compared to } j\text{th} \end{cases}$$

Then...

Expectation Analysis: Randomized Quicksort

Claim: $E[X_{i,j}] = \frac{2}{j-i+1}$ **Induction:** Assume true for all inputs of $< n$





Expectation Analysis: Randomized Quicksort

$$E[X] = \sum_{i=1}^n \sum_{j=i+1}^n E[X_{ij}]$$

$$E[X_{ij}] = \frac{2}{j - i + 1}$$

Expectation Analysis: Randomized Quicksort

$$E[X] = \sum_{i=1}^n \sum_{j=i+1}^n E[X_{ij}] \qquad E[X_{ij}] = \frac{2}{j-i+1}$$

$$E[X] = \sum_{i=1}^n 2\left(\frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n-i+1}\right) \leq 2n \log n$$

Note:

$$H_{n-i+1} - 1 = \left(\frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n-i+1}\right)$$

$$(H_{n-i+1} - 1) \leq H_n \leq \log n$$



Expectation Analysis: Randomized Quicksort

Summary: Randomized quick sort is $O(n \log n)$ regardless of input

Randomness:

Assumptions:



Probabilistic Accuracy: Fermat primality test

Pick a random a in the range $[2, p - 2]$

If p is prime and a is not divisible by p , then $a^{p-1} \equiv 1 \pmod{p}$

But... **sometimes** if n is composite and $a^{n-1} \equiv 1 \pmod{n}$



Probabilistic Accuracy: Fermat primality test

	$a^{p-1} \equiv 1(mod p)$	$a^{p-1} \not\equiv 1(mod p)$
p is prime		
p is not prime		



Probabilistic Accuracy: Fermat primality test

Let's assume $\alpha = .5$

First trial: $a = a_0$ and prime test returns 'prime!'

Second trial: $a = a_1$ and prime test returns 'prime!'

Third trial: $a = a_2$ and prime test returns 'not prime!'

Is our number prime?

What is our **false positive** probability? Our **false negative** probability?