

Applied Algorithms Assignment 1

Fei Fan (Peter) Chen

January 12, 2021

Exercise 1 (10 pts)

Using the fact that a fair coin can be constructed by assigning TH to be tails and HT to be heads, the expected number of tosses before we generate an unbiased coin flip is:

$$\frac{1}{\text{Prob}(\text{toss1} \neq \text{toss2})} \times 2 = \frac{1}{2pq} \times 2 = \frac{1}{pq}$$

Exercise 2 (10 pts)

Each randomized path has probability of $1/n^{(n-1)}$ since at each iteration, index j is taken from any index from $(0 \dots n)$. The number of permutations is only however $1/n!$. Since n^n is not divisible by $1/n!$ for $n > 2$ for the reason that n is not divisible by all the primes $p < n$, then so is $n^n - 1$. Therefore, there is always some permutation P that is reachable at the end of $n - 1$ iterations by more randomized paths than some other permutation P' . Therefore this algorithm cannot generate an uniform distribution for its possible permutations.

Exercise 3 (10 pts)

Part A

Three bits are needed for 6 numbers. We assign all 3-bit numbers less than 6 to 1..6 and start over for the remaining two 3-bit combinations.

$$E = 3/4 \times 3 + 1/4 \times (3 + E) = 4 \text{ bits} \leq 4 \text{ bits}$$

Since this is ≤ 4 bits, the next smallest number of bits that can represent 6 numbers will have expected bits greater than 4 since probability is always less than 1.

Part B

Four bits are needed for 9 numbers, We assign all 4-bit numbers less than 9 to 1...9 and start over for the remaining 7 4-bit combinations.

$$E = 9/16 \times 4 + 7/16 \times (4 + E) = 64/9 \text{ bits} \geq 4 \text{ bits}$$

There can be room for improvement all the way up to 8 bits...possibly. We try out the next bit size 5 (we map 3 combinations to each number).

$$E = 27/32 \times 5 + 5/32 \times (5 + E) = 160/32 \text{ bits} \leq 6 \text{ bits}$$

Exercise 4 (20 pts)

| N | Median-of-One | | Median-of-Three | | Median-of-Five | |
|----------|---------------|-------|-----------------|-------|----------------|-------|
| | Set | Pivot | Set | Pivot | Set | Pivot |
| 10 | 2.31 | 2.83 | 1.98 | 2.27 | 1.77 | 1.88 |
| 100 | 2.97 | 6.53 | 2.53 | 5.81 | 2.38 | 5.66 |
| 1000 | 3.38 | 11.77 | 2.84 | 10.12 | 2.48 | 9.17 |
| 10000 | 3.25 | 16.18 | 2.82 | 13.61 | 2.63 | 13.46 |
| 100000 | 3.37 | 20.97 | 2.71 | 17.41 | 2.55 | 16.79 |
| 1000000 | 3.15 | 24.65 | 2.72 | 21.85 | 2.54 | 20.51 |
| 10000000 | 3.44 | 30.48 | 2.78 | 25.82 | 2.57 | 24.05 |

Table 1: Average Case Analysis: average of 100 trials. Set Comparisons are normalized by N.

Observation 1: The number of comparisons seem to be around $2N$ to $3.5N$ as N increases from 10 to 10,000,000.

Observation 2: The number of pivots increases sub-linearly with N . In fact, the number of pivots increases at an additive constant rate of roughly 4 units for every $10x$ increase in N .

Observation 3: Both the number of pivots and number of comparisons in total improves with taking the median of more random pivots. As N increases, the improvement in number of total comparisons become more significant.

Code

```
import random
import statistics
import math

class QSelect():
    def __init__(self, medianN=1):
        self.medianN = medianN
        self.compSet = 0
        self.compPivot = 0

    def __call__(self, seq, k):
        # Randomly choose N elements from sequence
        medians = []
        for _ in range(self.medianN):
            medians.append(seq[int(random.uniform(0, len(seq)))])
        x = statistics.median(medians)

        # Create separate sets
```

```

s1 = [] # seq[i] < x
s2 = [] # seq[i] == x
s3 = [] # seq[i] > x
for value in seq:
    if value < x:
        s1.append(value)
    elif value == x:
        s3.append(value)
    else:
        s2.append(value)
self.compSet += len(seq)
if len(s2) >= k:
    self.compPivot += 1
    return self(s2, k)
elif len(s2) + len(s3) >= k:
    return x
else:
    self.compPivot += 1
    return self(s1, k - len(s2) - len(s3))

def reset(self):
    self.compSet = 0
    self.compPivot = 0

if __name__ == "__main__":
    random.Random(None)

    # Get average time complexity with 100 trials
    T = 100 # number of trials
    medianOfMedians = 5
    N = 10000000
    sequence = [i for i in range(N)]

    compSets = []
    compPivots = []

    q = QSelect(medianOfMedians)
    for _ in range(100):
        q.reset()
        q(sequence, int(math.ceil(N/2)))
        compSets.append(q.compSet/N)
        compPivots.append(q.compPivot)

    print( "mean compSets:", statistics.mean(compSets))
    print( "mean compPivots:", statistics.mean(compPivots))

```