

# Assignment: Statistical Principles

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CS 5140 Spring 2021

## Problem 1

```
# The size of the domain in question.  
n = 3000
```

## Part A

```
import random  
  
# Runs the experiment of generating random numbers in the range [0, n)  
# until two numbers generated equal. Returns the value `k` where the  
# k-th number equals to one of the numbers generated before.  
def experiment(n):  
    generated = set()  
    for k in range(1, n + 1):  
        x = random.randint(0, n - 1)  
        if x in generated: return k  
        generated.add(x)  
  
# Run the experiment once and report the k-value.  
k = experiment(n)  
print(f'Running the experiment once took {k} random trials')
```

Running the experiment once took 82 random trials

## Part B

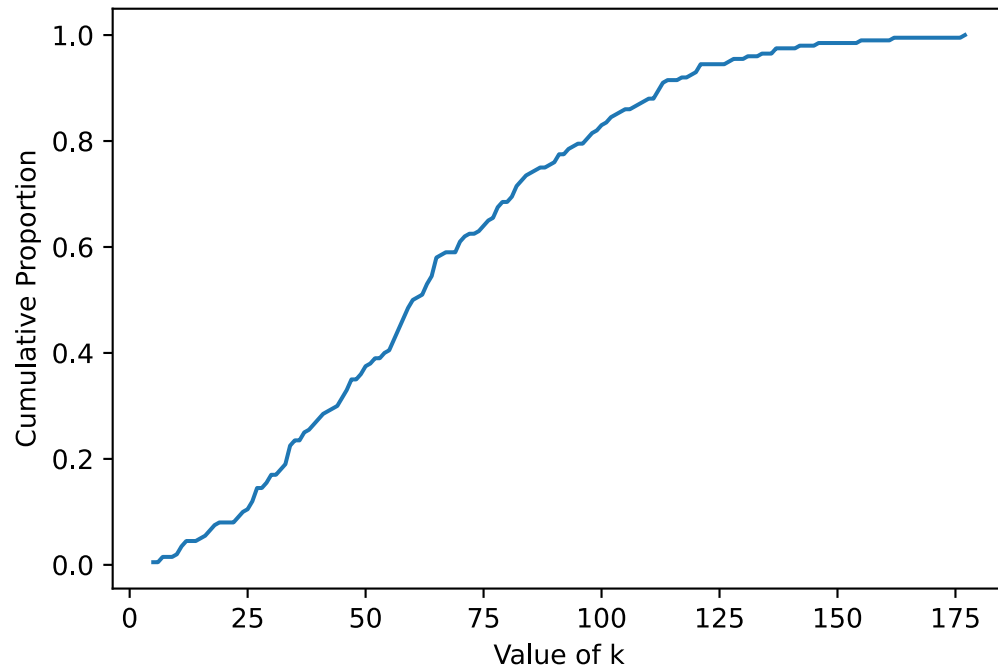
```
import itertools
from matplotlib import pyplot

# The number of experiments to run.
m = 200

# Run the experiment `m` times and record the frequency for each value
# of `k`.
k_freq = {}
for _ in range(m):
    k = experiment(n)
    k_freq[k] = k_freq.get(k, 0) + 1

# Plot the k-frequencies.
X = list(range(min(k_freq.keys()), max(k_freq.keys()) + 1))
Y = list(itertools.accumulate(k_freq.get(x, 0) / m for x in X))
pyplot.xlabel('Value of k')
pyplot.ylabel('Cumulative Proportion')
pyplot.plot(X, Y)
```

[<matplotlib.lines.Line2D at 0x23ee24ac088>]



## Part C

```
k_bar = sum(k * x for k, x in k_freq.items()) / m
print(f'Empirical estimate of the expected value of `k`: {k_bar}')
```

Empirical estimate of the expected value of `k`: 65.37

## Part D

In the implementation of the experiment:

- I used a hash-set to keep track of the numbers that had appeared (been randomly generated);
- I kept generating random numbers until the first time a number generated exists in that hash-set;
- When the above happened, I immediately terminated the experiment and reported the number of trials it took.

```

import time

# Returns the time in seconds that running an experiment with some
# values of `n` and `m` takes.
def time_experiment(n, m):
    start_time = time.time()
    while time.time() - start_time < .25: pass # warm-up loop
    start_time = time.time()
    for _ in range(m): experiment(n)
    return time.time() - start_time

t = time_experiment(n, m)
print(f'The experiment with n={n} and m={m} took {t:.3f} s')

```

The experiment with  $n=3000$  and  $m=200$  took 0.015 s

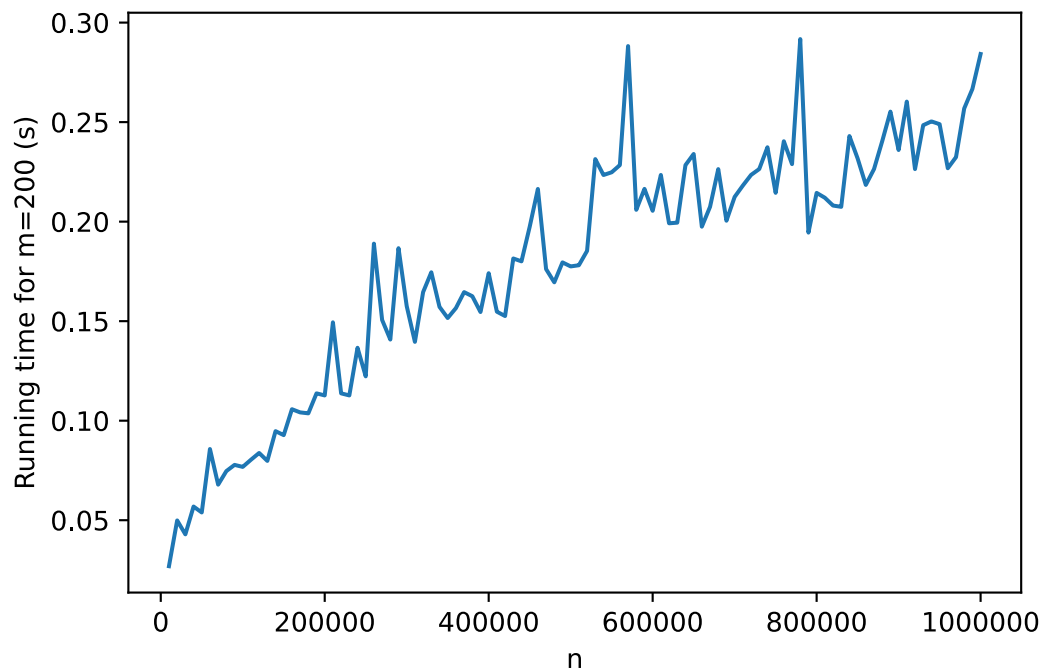
Now, for more values of  $n$  and  $m$ :

```

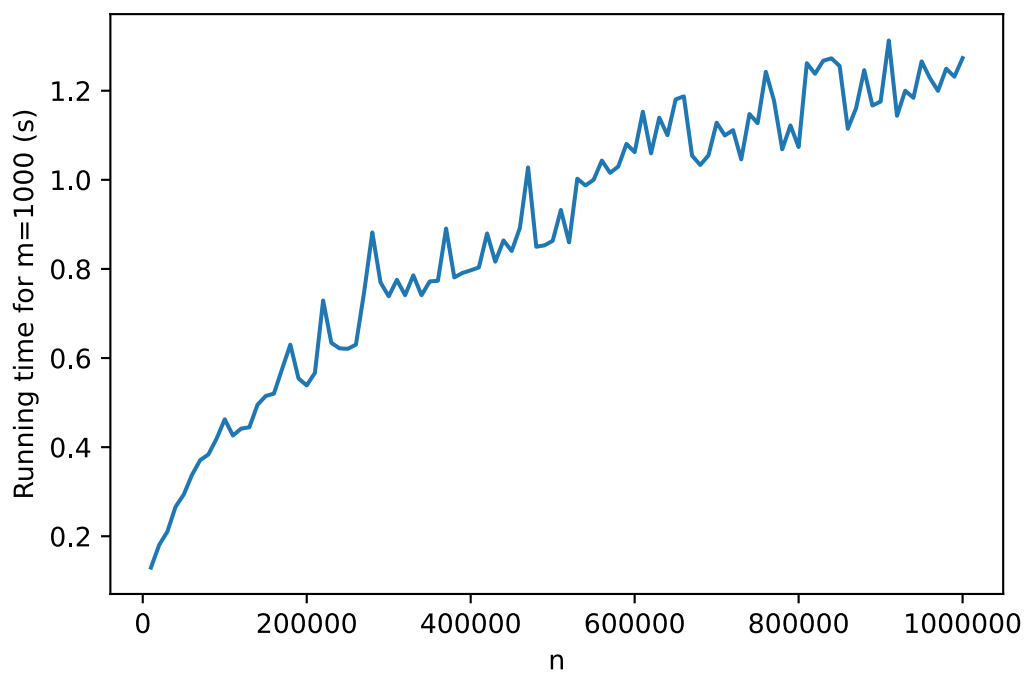
def time_m(m):
    X = list(range(10**4, 10**6 + 1, 10**4))
    Y = list(time_experiment(x, m) for x in X)
    pyplot.ticklabel_format(useOffset=False, style='plain')
    pyplot.xlabel('n')
    pyplot.ylabel(f'Running time for m={m} (s)')
    pyplot.plot(X, Y)

time_m(200)

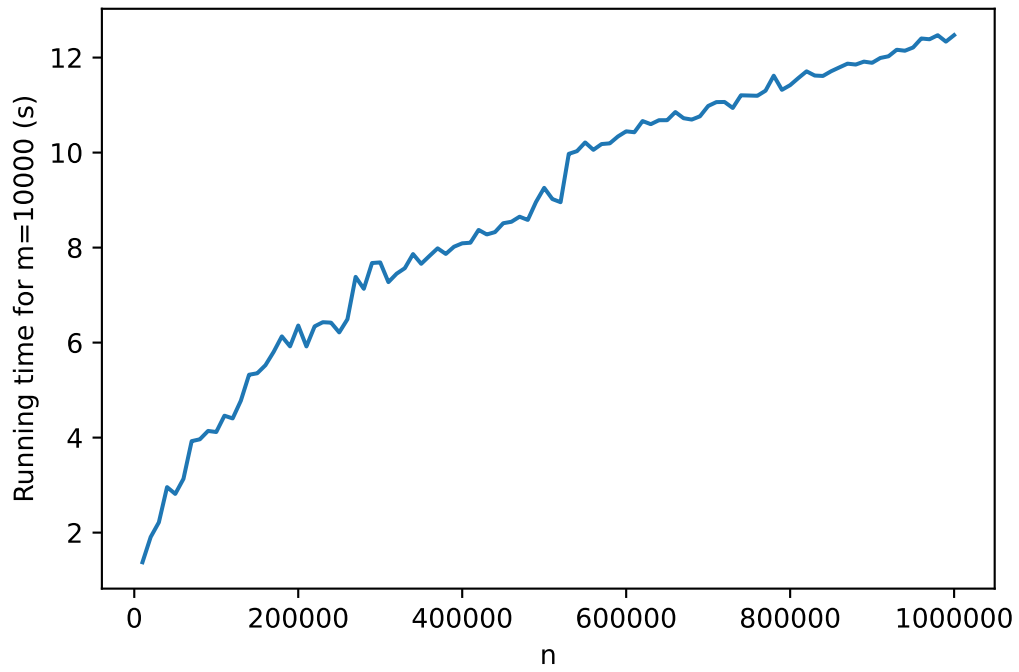
```



```
time_m(10**3)
```



```
time_m(10**4)
```



## Problem 2

```
# The size of the domain in question.  
n = 200
```

### Part A

```
import itertools, random  
  
# Runs the experiment of generating random numbers in the range [0, n)  
# until every possible number has been generated. Returns the value `k`  
# where as of the k-th generation, all possible numbers have been  
# generated.  
def experiment(n):  
    not_generated = {x for x in range(n)}  
    for k in itertools.count(1):  
        x = random.randint(0, n - 1)  
        not_generated.discard(x)  
        if not not_generated: return k  
  
# Run the experiment once and report the k-value.  
k = experiment(n)  
print(f'Running the experiment once took {k} random trials')
```

Running the experiment once took 1126 random trials

### Part B

```
from matplotlib import pyplot  
  
# The number of experiments to run.  
m = 300  
  
# Run the experiment `m` times and record the frequency for each value  
# of `k`.  
k_freq = {}
```

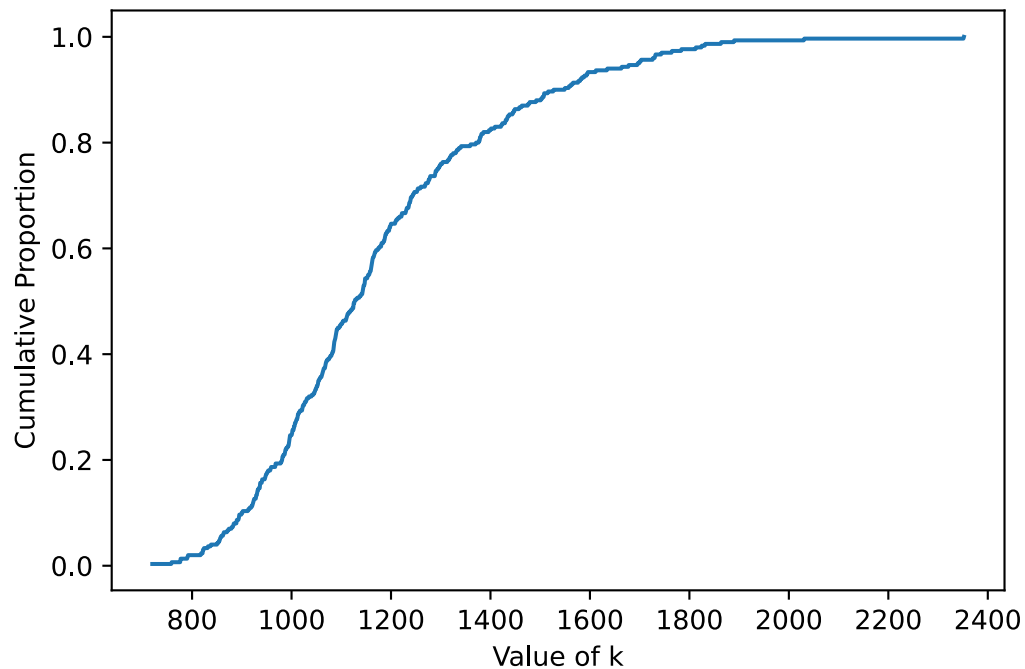
```

for _ in range(m):
    k = experiment(n)
    k_freq[k] = k_freq.get(k, 0) + 1

# Plot the k-frequencies.
X = list(range(min(k_freq.keys()), max(k_freq.keys()) + 1))
Y = list(itertools.accumulate(k_freq.get(x, 0) / m for x in X))
pyplot.xlabel('Value of k')
pyplot.ylabel('Cumulative Proportion')
pyplot.plot(X, Y)

```

[<matplotlib.lines.Line2D at 0x23ee2728d88>]





## Part C

```
k_bar = sum(k * x for k, x in k_freq.items()) / m
print(f'Empirical estimate of the expected value of `k`: {k_bar}')
```

Empirical estimate of the expected value of `k`: 1176.1733333333334

## Part D

In the implementation of the experiment:

- I used a hash-set to keep track of the numbers that had *not yet* appeared (*not* been randomly generated);
- I kept generating random numbers until that hash-set became empty, meaning that all possible numbers had been generated;
- When the above happened, I immediately terminated the experiment and reported the number of trails it took.

```
import time

# Returns the time in seconds that running an experiment with some
# values of `n` and `m` takes.
def time_experiment(n, m):
    start_time = time.time()
    while time.time() - start_time < .25: pass # warm-up loop
    start_time = time.time()
    for _ in range(m): experiment(n)
    return time.time() - start_time

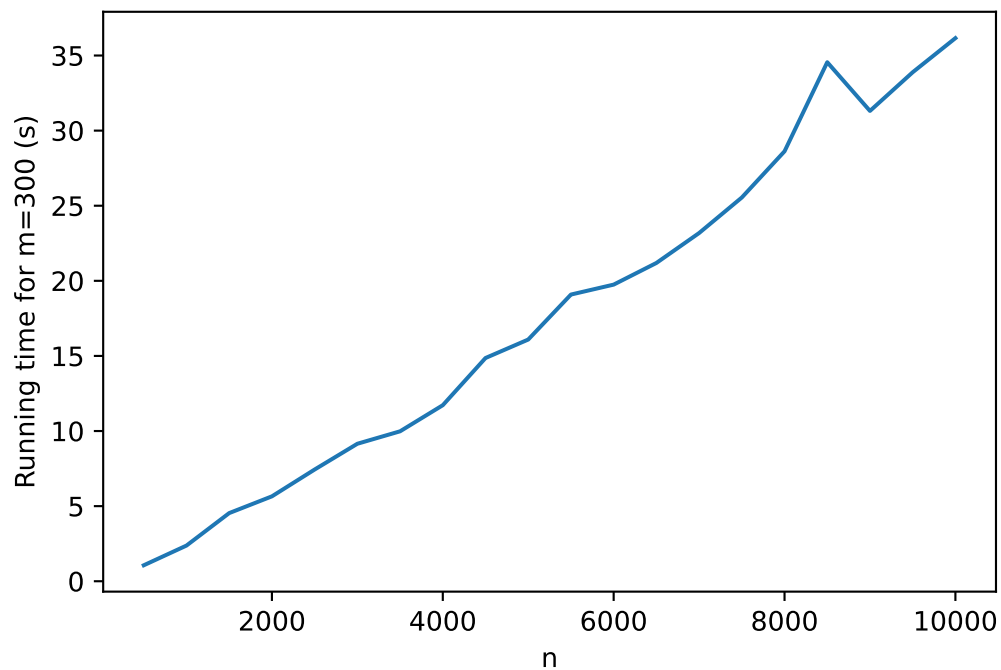
t = time_experiment(n, m)
print(f'The experiment with n={n} and m={m} took {t:.3f} s')
```

The experiment with  $n=200$  and  $m=300$  took 0.319 s

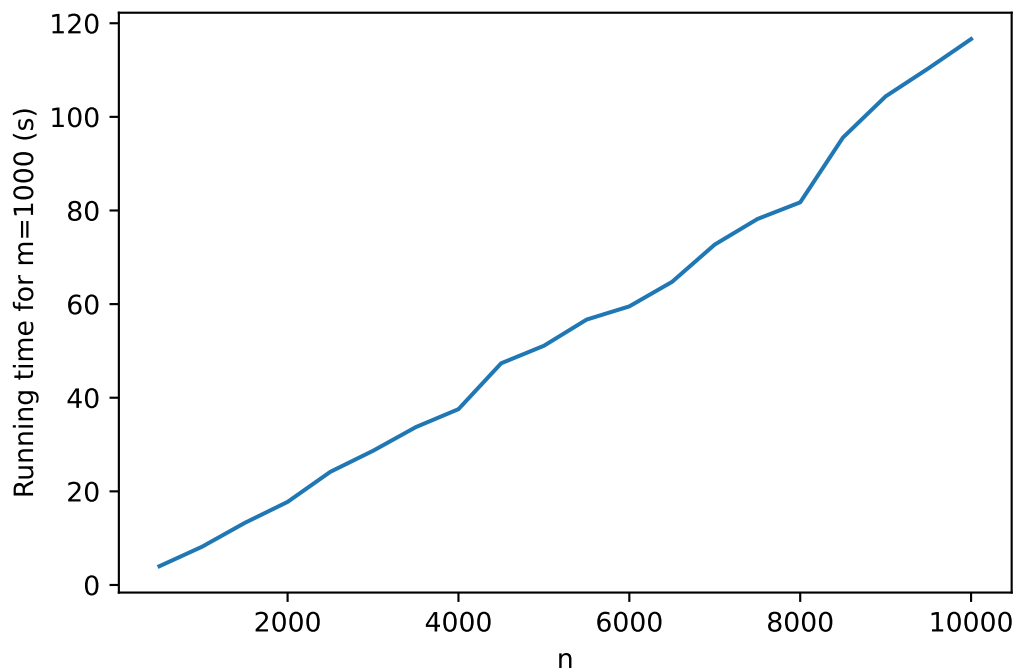
Now, for more values of  $n$  and  $m$ :

```
def time_m(m):
    X = list(range(500, 10**4 + 1, 500))
    Y = list(time_experiment(x, m) for x in X)
    pyplot.ticklabel_format(useOffset=False, style='plain')
    pyplot.xlabel('n')
    pyplot.ylabel(f'Running time for m={m} (s)')
    pyplot.plot(X, Y)
```

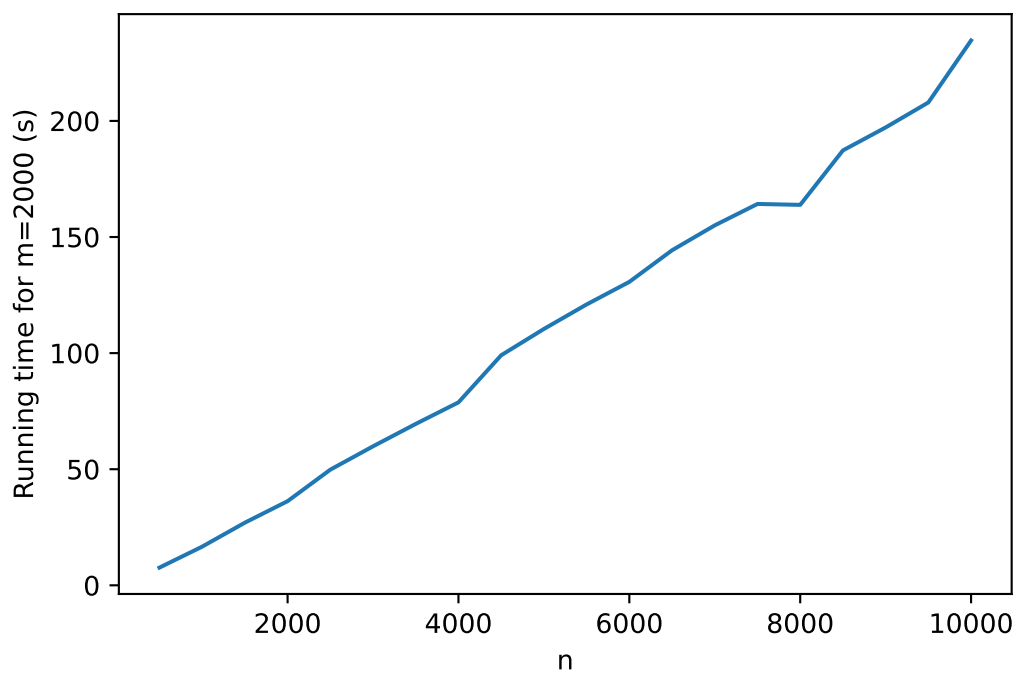
```
time_m(300)
```



```
time_m(1000)
```



```
time_m(2000)
```



## Problem 3

### Part A

I will use Method 1 from the lecture to estimate the actual expected value of  $k$ :

$$\begin{aligned}\Pr(\text{Collision with } k \text{ objects in a domain sized } n) &\approx 1 - \left(1 - \frac{1}{n}\right)^{k^2/2} \\ \frac{1}{2} &\approx 1 - \left(1 - \frac{1}{3000}\right)^{k^2/2} \\ \frac{k^2}{2} &\approx \log\left(\frac{1}{2}, \frac{2999}{3000}\right) \\ k &\approx \sqrt{2 \log\left(\frac{1}{2}, \frac{2999}{3000}\right)} \\ &\approx \boxed{64.48}\end{aligned}$$

Therefore, after generating about 65 random numbers, we can expect the probability that two of the random numbers equal to be more than 50%. This result is reasonably close to the estimate from *Problem 1-C* (65.37).

### Part B

I will use the (only) method from the lecture to estimate the actual expected value of  $k$ :

$$\begin{aligned}\mathbb{E}[k] &= n \cdot \sum_{i=1}^n \frac{1}{i} \\ &= 200 \cdot \sum_{i=1}^{200} \frac{1}{i} \\ &= 200 \cdot \left(1 + \frac{1}{2} + \cdots + \frac{1}{200}\right) \\ &\approx \boxed{1176}\end{aligned}$$

Therefore, we can expect to see every possibility being generated after generating about 1176 random numbers, which is reasonably close to the estimate from *Problem 2-C* (1176).

## Problem 4

Assume that the random numbers are generated from the domain  $\{1, \dots, n\}$ , which has size  $n$ . Now,  $f_i$  represents the number of trials that generate the number  $i$ , where  $1 \leq i \leq n$ .

$$\begin{aligned} \Pr \left[ \left| \mu - \frac{1}{n} \right| \geq \varepsilon \right] &= \Pr \left[ \exists i \in \{1, \dots, n\}, \left| \frac{f_i}{k} - \frac{1}{n} \right| \geq \varepsilon \right] \\ &= \Pr \left[ \left( \left| \frac{f_1}{k} - \frac{1}{n} \right| \geq \varepsilon \right) \cup \dots \cup \left( \left| \frac{f_n}{k} - \frac{1}{n} \right| \geq \varepsilon \right) \right] \\ &\leq \Pr \left[ \left| \frac{f_1}{k} - \frac{1}{n} \right| \geq \varepsilon \right] + \dots + \Pr \left[ \left| \frac{f_n}{k} - \frac{1}{n} \right| \geq \varepsilon \right] \end{aligned}$$

The above is true because the events  $|f_i/k - 1/n| \geq \varepsilon$  for any  $i$  are not disjoint, meaning that there could be more than one values of  $i$  where such an event is true. In other words, there could be more than one numbers in  $\{1, \dots, n\}$  that occur more often than  $\varepsilon + 1/n$ .

Suppose that  $\Pr[|f_i/k - 1/n| \geq \varepsilon] \leq .02/n$  for all  $i$ . This would directly imply that

$$\Pr \left[ \left| \frac{f_1}{k} - \frac{1}{n} \right| \geq \varepsilon \right] + \dots + \Pr \left[ \left| \frac{f_n}{k} - \frac{1}{n} \right| \geq \varepsilon \right] \leq n \cdot \frac{.02}{n} = .02$$

And thus,  $\Pr[|\mu - 1/n| \geq \varepsilon] \leq .02$ .

Now, in order for  $\Pr[|f_i/k - 1/n| \geq \varepsilon] \leq .02/n$  to be true for any  $i$ , first, notice that

$$\Pr \left[ \left| \frac{f_i}{k} - \frac{1}{n} \right| \geq \varepsilon \right] = \Pr \left[ \left| f_i - \frac{k}{n} \right| \geq k\varepsilon \right]$$

Consider a Chebyshev's Inequality for the variable  $X = f_i$ ; specifically,

$$\begin{aligned} \mathbb{E}[X] &= \mathbb{E}[f_i] = \frac{k}{n}, \\ \text{Var}[X] &= \text{Var}[f_i] = \text{Var} \left[ \text{Binom} \left( k, \frac{1}{n} \right) \right] = k \cdot \frac{1}{n} \cdot \frac{n-1}{n} = \frac{k(n-1)}{n^2} \end{aligned}$$

Therefore,

$$\begin{aligned}
\Pr[|X - \mathbb{E}[X]| \geq \varepsilon] &\leq \frac{\text{Var}[X]}{\varepsilon^2} \\
\Pr\left[\left|f_i - \frac{k}{n}\right| \geq k\varepsilon\right] &\leq \frac{k(n-1)}{n^2(k\varepsilon)^2} \leq \frac{.02}{n} \\
\frac{n-1}{nk\varepsilon^2} &\leq .02 \\
k &\geq \boxed{\frac{n-1}{.02n\varepsilon^2}}
\end{aligned}$$

By similar reasoning,

$$\Pr\left[\left|\mu - \frac{1}{n}\right| \geq \varepsilon\right] \leq .002 \implies k \geq \frac{n-1}{.002n\varepsilon^2}$$