Massachusetts Institute of Technology
6.042J/18.062J: Mathematics for Computer Science
Professors Albert R. Meyer and Shafi Goldwasser

Handout **??**
December 14, 1999

Final Exam

Your Name: _____

Circle the name of your TA:

Mary	Kyle	Sarfraz
Albert	Mike	Yana
Alantha	Ryan	Victor

- This exam is **closed** book; no calculators either. A two-sided sheet of prepared notes is allowed. **Your answers may cite without proof any of the results from lecture or handouts.**
- There are xx problems totaling xx points. The parts of a problem count equally unless noted otherwise.
- Put your name on the top of **every** page.
- Write your solutions in the space provided. If you need more space, write on the back of the sheet containing the problem.

  You need not show your work: a correct short answer with no explanation will receive full credit. But part credit for incorrect answers can only be given if the work leading to the answer is clearly shown.
- GOOD LUCK!

DO NOT WRITE BELOW THIS LINE

Problem	Points	Grade	Grader
1	35		
2	10		
3	10		
4	15		
5	15		
6	15		
7	15		
8	20		
9	15		
10	20		
11	15		
12	15		
Total	200		

Problem 1. Injections Let $f : D \to D$ be a function from some nonempty set D to itself. In the following propositions, x and y are variable ranging over D and g is a variable ranging over functions from D to D. Circle all of the propositions that are equivalent to the propositions that f is an *injection*:

a) $x = y \vee f(x) \neq f(y)$
b) $x = y \implies f(x) = f(y)$
c) $x \neq y \implies f(x) \neq f(y)$
d) $f(x) = f(y) \implies x = y$
e) $\neg[\exists x \exists y (x \neq y \wedge f(x) = f(y))$
f) $\neg[\exists z \forall x (f(x) \neq z)$
g) $\exists g \forall x (g(f(x)) = x)$
h) $\exists g \forall x (f(g(x)) = x)$

Problem 2. Infinite Cardinality

We have proved that the infinite set of real numbers is "strictly bigger" than the infinite set of integers, and the set of integers is "strictly bigger" than any finite set.

(a) State precisely what it means to say a set A is "strictly bigger" than another set B.

Solution:
There is no injection from A to B. Alternatively, there is no bijection between A and B, but there is an injection from B to A. **endsoln**

(b) Give an example of a set that is strictly bigger than the set of real numbers.

Solution: $P(\mathbb{R})$. **endsoln**

Problem 3. Expectation

The blunder in the published probabilistic analysis comparing CISC to RISC architecture performance was the assumption that

$$\mathrm{E}\left[\frac{R}{C}\right] = \frac{\mathrm{E}[R]}{\mathrm{E}[C]}$$

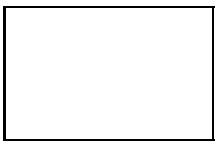
for random variables R and C. Give a simple example of two random variables R and C for which the assumption is not true.

Solution:

Let $R \equiv 1$, and $C ::= 1$ or 2 with equal probability. Then E[R]=1, E[C]=3/2, yield E[R]/E[C]=2/3. But E[R/C]=E[1/C]=3/4.

endsoln

Problem 4. Stars and Bars A juggler plans to juggle 11 rings for his finale. In his bag are 20 identical red rings, 30 identical blue rings, and 40 identical white rings. He must select at least one of each color to be patriotic. How many different patriotic choices of 11 rings in total are there for the act? An example of a patriotic choice of 11 rings would be: 3 red, 2 white, and 6 blue.



Solution:

This is the same as asking how many ways there are to arrange 11 stars and two bars such that there is at least one star in each of the three groups delimited by the bars. Reserving three stars for this purpose leaves 8 stars and 2 bars, so there are

$$\binom{8}{2}.$$

Note that the initial number of red, white, and blue rings available to the juggler is irrelevant as long as there are at least nine of each. **endsoln**

Problem 5. Modular Arithmetic

(a) Prove that there is no x such that

$$117 \cdot x \equiv 1 \pmod{213}.$$

Solution:
$117 \cdot x \equiv 1 \pmod{213}$ iff $117 \cdot x + 213 \cdot y = 1$ for some integer y. This is equivalent to having $\gcd(213, 117) = 1$, but by Euclid's algorithm, $\gcd(213, 117) = 3$.
endsoln

(b) Find an integer x for which

$$123 \cdot x \equiv 1 \pmod{2347} \quad _____$$

Solution:
$x = -706$ works because $1 = 37 \cdot 2347 + (-706) \cdot 123$. **endsoln**

Problem 6. Define a relation ρ on real-valued random variables X, Y by the rule that $X \ \rho \ Y$ iff $\Pr\{X \neq Y\} = 0$. Prove that ρ is an equivalence relation on random variables.

Solution:

ρ is *reflexive* because the event $X \neq X$ is the empty event—which has probability zero.

ρ is *symmetric* because the event $X \neq Y$ is the same as the event $Y \neq X$, so if either of these events has probability zero, then they both do.

To prove ρ is *transitive*, suppose $X \ \rho \ Y$ and $Y \ \rho \ Z$. We must show that if $X(s) \neq Z(s)$ for some sample point s, then $\Pr\{s\} = 0$. This implies $\Pr\{X \neq Z\} = 0$, so $X \ \rho \ Z$ as required.

So suppose $X(s) \neq Z(s)$. If also $X(s) \neq Y(s)$, then $\Pr\{(\}s) = 0$ because $X \rho Y$. Otherwise, $Y(s) = X(s) \neq Z(s)$, so $Y(s) \neq Z(s)$, and hence $\Pr\{s\} = 0$ in this case as well because $Y \ \rho \ Z$.

endsoln

Problem 7. Counting with Repetition

(a) How many distinguishable ways are there to rearrange the letters in

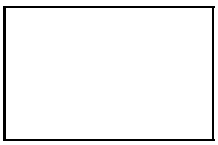
"**INSTRUCTOR**"?

Solution: There are 10! ways to arrange 10 distinct letters, but we have two T's and two R's, so this overcounts by a factor of 2! for the T's and by a factor of 2! for the R's. The correct answer is

$$\frac{10!}{2!2!} = 907200$$

endsoln

(b) ... with all vowels consecutive?



Solution:

Let the letter V represent the block of vowels, treated as a single unit. How many ways can we rearrange the letters NSTRCTRV? There are 8! ways to arrange 8 distinct letters, but we have two T's and two R's, so this overcounts by a factor of 2!2!. So there are $\frac{8!}{2!2!}$ ways to arrange the letters NSTRCTRV. There are three vowels, so there are 3! ways to arrange the vowels within the block V, so the number of ways to arrange the letters of INSTRUCTOR with all the vowels consecutive is

$$3!\frac{8!}{2!2!} = 60480$$

(Alternatively: $\frac{7!}{2!2!}$ ways to arrange the consonants, and 8 places to insert the 3 vowels within these 7 consonants.)

endsoln

Regular Problems

Problem 8. Linear Recurrence

You just moved into your new house and you are throwing a housewarming party. You invite 5 people on the first day and tell them to invite their friends. Each newly invited person invites 2 people the day after they are invited.

(a) Give a recurrence describing the number of people invited by the nth day.

Solution: You know that the number of people for each day will equal the number of people who were invited from the day before in addition to 2 new guests for each of the new people on the day before. T(n) = T(n-1) + 2 [T(n-1) - T(n-2)] = 3 T(n-1) - 2 T(n-2) **endsoln**

(b) Describe initial conditions for the recurrence.

Solution: You know that T(0) = 0 since no one was invited on day 0 and T(1) = 5, since you invited 5 people on the first day. **endsoln**

(c) Give a closed form expression equal to the number of people invited by the nth day.

Solution: Solving this recurrence $a^2 = 3a - 2$ $a^2 - 3a + 2 = 0$ $(a-2)(a-1) = 0$ $a = 2$ or 1

$T(n) = c2^n + b1^n = c2^n + b$

Plugging in the initial conditions, gives you two equations : $T(0) = 0 = c + b$ $T(1) = 5 = c * 2 + b$ $c = 5, b = -5$

$T(n) = (5 * 2^n) - 5$

endsoln

Problem 9. Binomial Identity. Suppose you want to choose a team of m people from a pool of n people for your startup company, and from these m people you want to choose k to be the team managers. Prove that you can do this in

$$\binom{n}{k}\binom{n-k}{m-k}$$

ways.

Solution: First, brute-force: To choose m out of n and then k out of chosen m, can be done in $\binom{n}{m}\binom{m}{k}$ ways. Now,

$$\begin{aligned}
\binom{n}{m}\binom{m}{k} &= \frac{n!}{m!(n-m)!}\frac{m!}{k!(m-k)!} \\
&= \frac{n!}{(n-m)!k!(m-k)!} \\
&= \frac{n!(n-k)!}{(n-m)!k!(m-k)!(n-k)!} \\
&= \frac{n!}{k!(n-k)!}\frac{(n-k)!}{(n-m)!(m-k)!} \\
&= \frac{n!}{k!(n-k)!}\frac{(n-k)!}{((n-k)-(m-k))!(m-k)!} \\
&= \binom{n}{k}\binom{n-k}{m-k}
\end{aligned}$$

Another way to look at this. Instead of choosing first m from n and then k from m, you could alternately choose the k managers from the n people and then choose $m-k$ people to fill out the team from the remaining $n-k$ people. This gives you $\binom{n}{k}\binom{n-k}{m-k}$ ways of picking your team. Since you must have the same number of options regardless of which order you choose to pick team members and managers in, $\binom{n}{m}\binom{m}{k} = \binom{n}{k}\binom{n-k}{m-k}$. **endsoln**

Problem 10. Trees and Structural Induction A *full binary tree* with its *vertices* and *leaves* is defined inductively as follows:

- A simple graph with one vertex is a full binary tree. The vertex is called the *root* of the tree. The *set of leaves* of the tree is the set whose sole member is the root. The *set of vertices* of this tree is the same as the set of leaves.
- If B_1 and B_2 are full binary trees such that the vertices of B_1 and B_2 are disjoint, then a simple graph consisting of a new vertex with edges from the new vertex to the roots of B_1 and B_2 is also a full binary tree, B. The *root of* B is the new vertex. The *vertices of* B are exactly its root, the vertices of B_1, and the vertices of B_2. The *leaves of* B are exactly the leaves and B_1 and the leaves of B_2.

For example, a full binary tree with 5 nodes and 3 leafs is is shown in Figure **INSERT FIGURE**.

Prove that for all full binary trees B,

$$|\text{leaves}(B)| = \frac{|\text{vertices}(B)| + 1}{2}.$$

Solution: The induction hypothesis, $P(B)$, is the equation to be proved.

Base case P(single root): there is 1 leaf, 1 vertex and indeed $1 = (1+1)/2$.

Inductive step B has a root r connected to two full binary trees B_1, B_2; the induction hypothesis is that $P(B_1)$ and $P(B_2)$ are true. So

$$\begin{aligned}
|\text{leaves}(B)| &= |\text{leaves}(B_1) \cup \text{leaves}(B_2)| \\
&= |\text{leaves}(B_1)| + |\text{leaves}(B_2)| \qquad \text{because vertices of } B_1 \text{ and } B_2 \text{ are disjoint} \\
&= \frac{|\text{vertices}(B_1)| + 1}{2} + \frac{|\text{vertices}(B_2)| + 1}{2} \qquad \text{by induction hypothesis} \\
&= 1 + \frac{|\text{vertices}(B_1)| + |\text{vertices}(B_2)|}{2} \\
&= |\{r\} \cup \text{vertices}(B_1) \cup \text{vertices}(B_2)| \qquad \text{disjointness again} \\
&= |\text{vertices}(B)|
\end{aligned}$$

endsoln

Problem 11. Conditional Probability.

Suppose that a resident of Boston chosen at random is a Celtics fan with probability 0.90, that a resident chosen at random is a Red Sox fan with probability 0.75, and that a resident Celtcis fan chosen at random is a Red Sox fan with probability 0.80. Now suppose that a resident is chosen at random from among those that are *not* Celtics fans. What is the probability that such a resident is a Red Sox fan?

Solution: Let P represent the event that a resident is a Red Sox fan, and S represent the event that the resident is a Celtics fan. Then:

$$\begin{aligned}
\Pr\{P\} &= \Pr\{P \mid S\}\Pr\{S\} + \Pr\left\{P \mid \overline{S}\right\}\Pr\left\{\overline{S}\right\} \\
\Pr\left\{P \mid \overline{S}\right\} &= \frac{\Pr\{P\} - \Pr\{P \mid S\} * \Pr\{S\}}{\Pr\left\{\overline{S}\right\}} \\
&= \frac{0.75 - 0.80 \cdot 0.90}{1 - 0.9} \\
&= 0.3
\end{aligned}$$

endsoln

Problem 12. Chebychev Bound

A man has a pocketful of n keys, exactly one of will open the door to his apartment. He picks a random key from his pocket, tries to open the door with it, and puts it back in his pocket if it doesn't work (no, he isn't very together about key management). He continues until he finds the key that works.

(a) What is the expected number of keys the man draws until he finds the right one? *Hint*: Mean time to failure.

Solution: The probability of picking the right key on each trial is $1/n$, so mean time to failure is $1/(1/n) = n$. **endsoln**

(b) What is the variance for the number of keys he picks? *Hint*: Some relevant summation formulas are in the Appendix.

Solution: The probability stays the same on all subsequent trials since the picked key is always returned to the urn. Thus

$$\Pr\{T = k\} = \left(\frac{n-1}{n}\right)^{k-1} \frac{1}{n}.$$

Note: this is the same as the waiting time for the first head in the Bernouli process with a biased coin. Let $p = 1/n$ and let $q = 1 - p$.
The variance of T can be computed directly from the formula,

$$\mathrm{Var}[T] = \mathrm{E}\left[T^2\right] - \mathrm{E}^2[T].$$

Again, by definition we have

$$\mathrm{E}\left[T^2\right] = \sum_{k=1}^{\infty} k^2 p q^{k-1}.$$

We know that

$$\sum_{k=1}^{\infty} k q^k = q \sum_{k=1}^{\infty} k q^{k-1} = \frac{q}{(1-q)^2}.$$

Moreover,

$$\begin{aligned}
\sum_{k=1}^{\infty} k^2 q^{k-1} &= \sum_{k=1}^{\infty} \frac{d\, k q^k}{dq} \\
&= 1/q \sum_{k=1}^{\infty} k q^k \\
&= 1/q \frac{q}{(1-q)^2} \\
&= \frac{1}{(1-q)^2} + \frac{2q}{(1-q)^3} = \frac{1}{p^2} + \frac{2q}{p^3}.
\end{aligned}$$

Finally, the expectation $\mathrm{E}\left[T^2\right]$ is

$$\mathrm{E}\left[T^2\right] = p\left(\frac{1}{p^2} + \frac{2q}{p^3}\right) = \frac{1}{p} + \frac{2q}{p^2},$$

and the variance is

$$\mathrm{Var}\left[T\right] = 1/p + \frac{2q}{p^2} - \frac{1}{p^2} = \frac{q}{p^2} = n(n-1).$$

endsoln

(c) What is the Chebyshev upper bound on the probability that the man will have to pick at least 100 keys? Let V denote the variance that you were asked to calculate in the previous part, and express your answer as a closed formula using V.

Solution:
One-way bound:

$$\Pr\{K \geq 100\} = \Pr\{K - n \geq 100 - n\} \leq \frac{V}{V + (100-n)^2}.$$

Also OK to use the regular two-way bound.
endsoln

Problem 13. Markov's Bounds

A herd of 350 cows is stricken by an outbreak of "mad cow" disease. The disease lowers the normal body temperature of a cow, and a cow will die if its temperature goes below 90 degrees F. The disease epidemic is so intense that it lowered the average temperature of the cows to 45 degrees. The lowest temperature measured was 20 degrees.

(a) State a best possible upper bound on the number of cows whose temperature is high enough to survive.

Solution:
Let R be the temperature. Apply Markov's Bound to $R - 20$:

$$\Pr\{R \geq 90\} = \Pr\{R - 20 \geq 70\} \leq \mathrm{E}[R - 20]/70 = (45 - 20)/70 = 5/14,$$

so at most $(5/14)350 =$ **125** cows have temperature ≥ 90 degrees.
endsoln

(b) Give an example set of temperatures that achieve your bound.

Solution:
Of the 350 cows, 125 have 90 degree temperature and 225 have 20 degree temperature. The mean is

$$(125 \cdot 90 + 225 \cdot 20)/350 = 45,$$

as required. **endsoln**

Problem 14. Random Graphs A simple graph with 10 vertices is constructed by randomly placing an edge between every two vertices with probability p. These random edge placements are performed independently.

(a) What is the probability that a given vertex, v, of the graph has degree two?

Solution: $\binom{9}{2} \cdot p^2 \cdot (p-1)^7$. **endsoln**

(b) What is the expected number of nodes with degree two?

Solution: 10 times the previous answer by linearity of expectations. **endsoln**

(c) What is the expected number of cliques of size k, that is, subgraphs isomorphic to the complete graph with k vertices? Express your answer as a closed form formula involving binomial coefficients.

Solution:
The strategy for computing the expected number of cliques is again: linearity of expectation. Define an indicator random variable for every possible k-clique. Every k-subset of the n nodes is a possible k-clique, so there are $\binom{n}{k}$ such random variables. The random variable for a particular k-clique is 1 if the k-clique is found in the graph and 0 if not, so its expected value is just the probability that a particular k-clique occurs in the graph. This happens if and only if all k nodes are connected by edges; there are $\binom{k}{2}$ such edges, so the probability of a particular k-clique is $p^{\binom{k}{2}}$. Summing the expected values of all the k-cliques gives the expected number of k-cliques:

$$\binom{n}{k} p^{\binom{k}{2}}.$$

endsoln

(d) The previous result impliies an interesting fact which you may assume without proof: if $p = 1/2$, then the expected number of cliques of size $2 \log n$ is greater than one. Oddly, even though we can expect there to be such a clique, no procedure is known to *find* some clique of size $2 \log n$ in the graph—when there is one—that is much more efficient than exhaustively searching through all possible subsets of vertices of size $2 \log n$ and checking each subset to see if it is a clique. Circle each of the following expressions which correctly bounds the number of subsets such an exhaustive search would examine:

a) $O(2^n)$
b) $O(n^2)$
c) $O(n^{log\, n})$
d) $\Theta(e^{\log^2 n})$
e) $\Theta((\log n)^n)$

Solution: $O(n^{log\, n})$, and $O(2^n)$. **endsoln**

Problem 15. We have three coins to flip: one has probability 3/4 of coming up heads, another has probability 7/8 of coming up heads, and finally there is one with probability 15/16 of coming up heads. One of the three coins is picked, and this coin is repeatedly tossed.

Explain a way to determine a number of tosses sufficient to allow identification with 99% confidence of the coin being tossed. You do not have to carry out the calculation, and you may cite without proof any of the results from the Appendix or other course material.

Solution:

Identify the coin to be the one whose probability is closest to the fraction of heads among n flips. This identification will be correct 99% of the time if the probability that the actual fraction of heads in n flips differs from the expected fraction of heads by more than 1/32 in at most 0.01. We choose 1/32 because it is half the distance between the two closest coin probabilities.

Let S_n be the number of heads appearing in n flips, so S_n/n is the fraction of heads. We assume the successive flips are independent, so the variance of S_n/n is $1/n^2$ times the variance of S_n which is n times the variance of a Bernoulli variable whose probability was the probability of heads in one flip. We know the variance of a Bernoulli variable is at most 1/4, so $\mathrm{Var}\,[S_n/n] = (1/n^2)(1/4)n = 1/4n$. Chebyshev's bound implies that the probability that S_n/n deviates from its mean by 1/32 is at most $(1/4n)/(1/32)^2$. Choose n so this quantity is at most 0.01.

Alternatively, observe that S_n will have a Bernoulli distribution with one of the three probabilities 3/4, 7/8, 15/16. Using the bounds from Lecture 21 on the tails of the binomial distribution, we could choose n large enough that the probability that $\Pr\{|S_n - pn| \geq n/32\} \leq 0.01$ for each $p = 3/4, 7/8$, and 15/16.

endsoln

Appendix

Graphs

A **simple graph** is an undirected graph with no self loops and at most one edge between any two nodes.

A **tree** is a connected, acyclic simple graph.

An **Eulerian circuit** in a graph is a path beginning and ending at the same vertex and containing each edge exactly once.

The **complete graph**, K_n, is the simple graph with n vertices and with an edge between every two vertices.

The **degree** of a vertex in a graph is the number of edges at the vertex.

A *connected component* of a simple graph is an equivalence class of vertices under the "is connected by a path to" relation on pairs of vertices.

Partial Orders

Lexicographic order, $\preceq$, on pairs of natural numbers is defined by the rule that

$$(n_1, n_2) \preceq (m_1, m_2) \quad \text{iff} \quad [(n_1 < m_1) \text{ or } (n_1 = m_1 \text{ and } n_2 \leq m_2)].$$

Sums

$$\sum_{i=0}^{n} r^i = \frac{r^{n+1} - 1}{r - 1}.$$

$$\sum_{i=0}^{n} i r^i = \frac{r - (n+1)r^{n+1} + n r^{n+2}}{(1-r)^2}.$$

$$\int \ln x \, dx = x \ln(x/e).$$

0.1 Combinatorics

Stirling's formula:

$$n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n.$$

The Principle of Inclusion-Exclusion:

$$|A \cup B| = |A| + |B| - |A \cap B| .$$

For n sets:

$$\left| \bigcup_{i=1}^{n} A_i \right| = \sum_{S \subset \{1,\ldots,n\}} (-1)^{|S|} \left| \bigcap_{j \in S} A_j \right| .$$

Probabilistic version:

$$\Pr\{A \cup B\} = \Pr\{A\} + \Pr\{B\} - \Pr\{A \cap B\} .$$

$$\Pr\left\{ \bigcup_{i=1}^{n} A_i \right\} = \sum_{S \subset \{1,\ldots,n\}} (-1)^{|S|} \Pr\left\{ \bigcap_{j \in S} A_j \right\} .$$

$$\begin{aligned} n! &::= n \cdot (n-1) \cdots 2 \cdot 1 \\ 0! &::= 1 \\ P(n,r) &::= n \cdot (n-1) \cdots (n-(r-2) \cdot (n-(r-1)) \\ C(n,r) = \binom{n}{r} &::= \frac{n!}{r!(n-r)!} \end{aligned}$$

Markov & Chebyshev Bounds

Markov bound:

$$\Pr\{R \geq x\} \leq \frac{\mathrm{E}[R]}{x}.$$

Chebyshev bound: for $c \geq 0$,

$$\Pr\{|R - \mathrm{E}[R]| \geq c\sigma_R\} \leq \frac{1}{c^2}.$$

The "one-way" version of Chebyshev bound: for $c \geq 0$,

$$\Pr\{R - \mathrm{E}[R] \geq c\sigma_R\} \leq \frac{1}{1+c^2}.$$

Weak Law of Large Numbers for Bernoulli Variables

A *Bernoulli variable* is a 0-1-valued random variable. For any event A, the *index variable*, I_A, is the Bernoulli variable which is 1 exactly when event A occurs.

Let $G_1, G_2, \ldots, G_n$ independent Bernoulli variables, each with the same probability p of being equal to one. In this case, $G_i = G_i^2$, so

$$\mathrm{E}\left[G_i^2\right] = \mathrm{E}\left[G_i\right] = \Pr\left\{G_i = 1\right\} = p,$$

and

$$\mathrm{Var}\left[G_i\right] = \mathrm{E}\left[G_i^2\right] - \mathrm{E}^2\left[G_i\right] = p - p^2 = p(1-p).$$

Let

$$S_n ::= \sum_{i=1}^{n} G_i,$$

We can use the random variable S_n/n to estimate p. This estimator S_n/n is a random variable with two critical properties:

$$\begin{aligned} \mathrm{E}\left[\frac{S_n}{n}\right] &= \mathrm{E}\left[G_i\right] = p, \\ \mathrm{Var}\left[\frac{S_n}{n}\right] &= \frac{\mathrm{Var}\left[G_i\right]}{n} = \frac{p(1-p)}{n}. \end{aligned}$$

To prove this, note that by linearity of expectation

$$\mathrm{E}\left[\frac{S_n}{n}\right] = \frac{\mathrm{E}\left[\sum_{i=1}^{n} G_i\right]}{n} = \frac{\sum_{i=1}^{n} \mathrm{E}\left[G_i\right]}{n} = \frac{np}{n} = p.$$

Also, since the G_i's are independent, their variances will add, so

$$\begin{aligned} \mathrm{Var}\left[\frac{S_n}{n}\right] &= (\frac{1}{n})^2 \mathrm{Var}\left[S_n\right] \\ &= (\frac{1}{n})^2 \sum_{i=1}^{n} \mathrm{Var}\left[G_i\right] \\ &= (\frac{1}{n})^2 n \mathrm{Var}\left[G_i\right] \\ &= \frac{\mathrm{Var}\left[G\right]}{n}. \end{aligned}$$

Now Chebyshev's Bound tells us that

Theorem 0.1. *Let $S_n ::= \sum_{i=1}^{n} G_i$ where $G_1, \ldots, G_n$ are mutually independent Bernoulli variables with the same probability p of being equal to one. Then*

$$\Pr\left\{\left|\frac{S_n}{n} - p\right| \geq x\right\} \leq \frac{p(1-p)}{nx^2}.$$

Now if we fix p and x and let n approach infinity, we conclude that the Weak Law of Large Numbers holds:

$$\lim_{n \to \infty} \Pr \left\{ \left| \frac{S_n}{n} - p \right| \geq x \right\} = 0.$$