**Maximum score:** 145 points.

Instructions: For this test, you work in teams to solve multi-part, proof-oriented questions. Problems that use the words "compute," "find," "draw," or "write" require only an answer; no explanation or proof is needed. Unless otherwise stated, all other questions require explanation or proof. The problems are ordered by content, *not difficulty*. The difficulties of the problems are generally indicated by the point values assigned to them; it is to your advantage to attempt problems throughout the test. In your solution for a given problem, you may cite the statements of earlier problems (but not later ones) without additional justification, even if you haven't solved them. Footnotes are not necessary to understand or solve the contents of the round.

1 Graphs and Proofs (45 pts)

Welcome to the power round! We will learn a new form of proof, which you probably haven't heard of. It is a type of proof that lies at the center of modern cryptography and security–the **zero-knowledge proof** (or ZKP).

To begin, let's give some background on problems that ZKPs solve. Consider a group of people at an event, like a math contest, who are friends with some of the other people at the event. If we wanted to represent the relations between these people, we could use a point to represent each person and lines between points to denote that the people represented by the points in question are friends. This makes sense because friendships are symmetric relationships (if A is friends with B, then B is friends with A), and no one is friends with themselves. Such a representation has a name.

> **Definition 1.1.** A **graph** $G = (V, E)$ is a set of vertices, V, and a set of edges, E. Each edge is itself an (unordered) set of two distinct vertices. Vertices that share an edge between them are called **adjacent**. The number of vertices in a graph is denoted $|V|$ and the number of edges is denoted $|E|$.
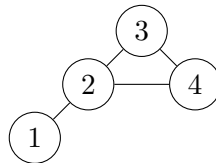
In the example above, vertices could represent people at our event, while an edge could indicate that the two people sharing the edge are friends. Formally, one can record this graph with vertex set $V = \{1, 2, 3, 4\}$ and edge set $E = \{\{1, 2\}, \{2, 3\}, \{3, 4\}, \{2, 4\}\}$.
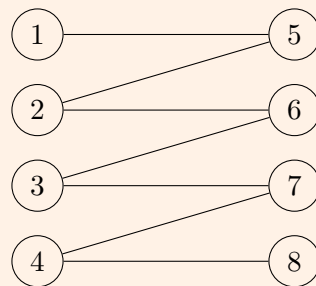
Question 1.1 (2 pts). Draw a graph with $|V| = 5$ and $|E| = 7$ and give its vertex and edge sets.

Solution: There are many possible solutions to this problem. One example is the following graph:
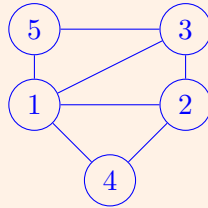
Its vertex set is $V = \{1, 2, 3, 4, 5\}$ and its edge set is
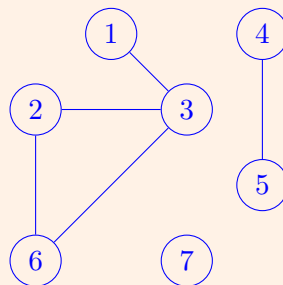$E = \{\{1, 2\}, \{2, 3\}, \{3, 5\}, \{2, 4\}, \{1, 5\}, \{1, 3\}, \{1, 4\}\}$.

Question 1.2 (2 pts). Find the vertex and edge sets of the following graph.

Solution: The vertex set is $V = \{1, 2, 3, 4, 5, 6, 7, 8\}$. The edge set is
$E = \{\{1, 5\}, \{2, 5\}, \{2, 6\}, \{3, 6\}, \{3, 7\}, \{4, 7\}, \{4, 8\}\}$

Question 1.3 (2 pts). Draw the graph with the vertex set $V = \{1, 2, 3, 4, 5, 6, 7\}$ and edge set $E = \{\{2, 3\}, \{4, 5\}, \{1, 3\}, \{3, 6\}, \{2, 6\}\}$.

Solution:

Question 1.4 (3 pts). Suppose a graph has n vertices. Compute the lowest number and highest number of edges it could have.

**Solution**: The lowest number of edges possible is 0, since there is no obligation for a graph to have edges.
In a graph with the highest number of edges, each of the n vertices would have an edge to each of the other $n - 1$ vertices. In this case, there is one edge for each combination of two vertices, which is equal to $\binom{n}{2} = \frac{n(n-1)}{2}$

Graphs can be used for a variety of situations, meaning that different situations could result in similar graphs. For example, a graph to model three people who know each other could be very similar to a graph modeling three cities that all have direct routes to each other. Functionally, these graphs are the same, so we have an important distinction for them.

Definition 1.2. An **isomorphism** between graphs G_1 and G_2 is a function f that maps vertices from one graph to another so that edges in G_1 are also represented in G_2, and edges in G_2 all correspond to some edge of G_1. More precisely, f is an invertible function (also called a **bijection**) such that $\{u, v\}$ is an edge of G_1 if and only if $\{f(u), f(v)\}$ is an edge of G_2.
We say graphs G_1, G_2 are **isomorphic** and write $G_1 \cong G_2$ if there's an isomorphism between them.

Intuitively, an isomorphism between two graphs G_1, G_2 just means that one can re-label the vertices of G_1 such that the resulting graph is exactly the graph G_2. Here are two isomorphic graphs.
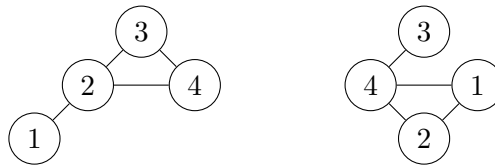
One isomorphism f to get from the left graph to the right graph is given by $f(1) = 3, f(2) = 4, f(3) = 1, f(4) = 2$.

Question 1.5 (4 pts). For the following eight graphs, find all isomorphic pairs. If you find two graphs that are isomorphic, give the isomorphism from one to the other (either order is fine).
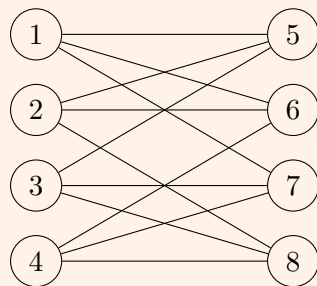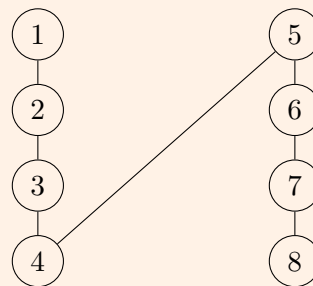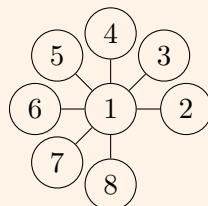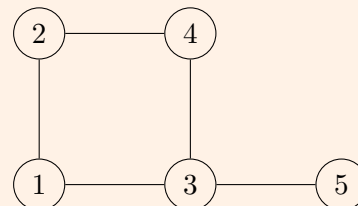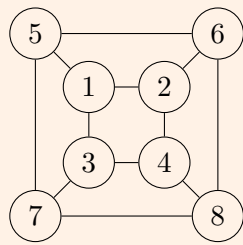
G_5
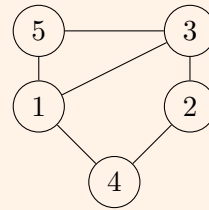
G_6
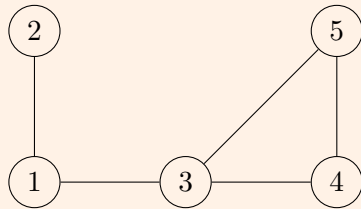
G_7
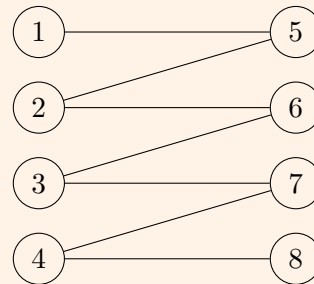
G_8

Solution: There may be more than one possible isomorphism for a pair of isomorphic graphs.
$G_2 \cong G_8$. One isomorphism f from G_2 to G_8 is defined as $f(1) = 1, f(2) = 5, f(3) = 2, f(4) = 6, f(5) = 3, f(6) = 7, f(7) = 4, f(8) = 8$
$G_1 \cong G_5$. One isomorphism f from G_1 to G_5 is defined as $f(1) = 1, f(2) = 4, f(3) = 6, f(4) = 7, f(5) = 2, f(6) = 3, f(7) = 5, f(8) = 8$

Question 1.6 (5 pts). If we consider isomorphic graphs to be the same, how many distinct graphs are there with four vertices?

Solution: There are 11 of them. If graphs have a different number of edges, they can't be isomorphic. Thus, we can count the number of distinct graphs with 0 to 6 edges (the minimum and maximum number of edges possible). All 4-vertex graphs with $0, 1, 5,$ or 6 are isomorphic to all other 4-vertex graphs with the same number of edges. If a graph has 2 edges, the edges either share a vertex or they don't, so there are 2 distinct graphs. Similarly, among 4-vertex graphs with 4 edges, the 2 "missing" edges (edges that would be present on the 6-edge graph) either share a vertex or don't, resulting in 2 distinct 4-edge graphs.

If there are 3 edges, they either form a cycle (such as $\{(1, 2), (2, 3), (1, 3)\}$), contain edges that can be ordered into a line (such as $\{(1, 2), (2, 3), (3, 4)\}$), or all share the same vertex (such as $\{(1, 2), (1, 3), (1, 4)\}$). All other graphs with 4 vertices and 3 edges are isomorphic to one of these. Thus, there are 3 distinct graphs with 3 edges, so there are 11 distinct graphs in total.

Question 1.7 (2 pts). Given two graphs G_1, G_2, both with n vertices, how many bijections are there between the vertex sets of the two graphs?

Solution: Each bijection corresponds to a unique permutation of the vertices, so there are $n!$ bijections.

Question 1.8 (5 pts). Draw two isomorphic graphs with exactly 6 vertices and with exactly one isomorphism between them.

Solution: Here is one example of graphs with only 1 isomorphism:
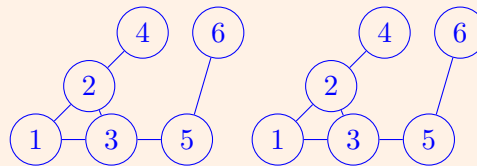
Let's now develop a deterministic procedure to check whether graphs are isomorphic to each other.

Question 1.9 (3 pts). Suppose someone hands you two graphs (e.g. their vertex and edge sets) $G_1 = (V_1, E_1)$ and $G_2 = (V_2, E_2)$ which both have n vertices and m edges. Devise a method to check if the two graphs are isomorphic.

Solution: There only exist $n!$ ways to assign the vertices of G_1 to G_2. Thus, we could check each of those assignments to determine whether it is a valid isomorphism. If one of those assignments is an isomorphism, then the graphs are isomorphic. Otherwise, the graphs are not isomorphic.

It might be that your method is not very fast to implement. Suppose I ask you to pay for your method (or *algorithm*).

Definition 1.3. The **cost** of an algorithm is defined as the sum of the costs of the basic operations required to implement the algorithm. For graphs, if you need to iterate through the elements of an edge set, it costs you \$1 per element. If you want to check if an element is in a set, this also costs you \$1. For instance, if I wanted to look at every edge in G_1, it would cost me \$$m$.

Solution: For the purposes of this round, we only care about the asymptotics, so we were a bit lenient with off-by-one errors.

Question 1.10 (4 pts). Compute the possible cost of your algorithm to Question 1.9 in the worst case. Your answer may depend on m and n. Don't worry about making the cost of the algorithm as small as possible; a correct algorithm and correct cost analysis is all we need.

Solution: For each of the $n!$ possible assignments of vertices from G_1 to G_2, each of the m edges needs to be checked to determine whether an isomorphism has been found. Thus, the total cost is $mn!$. For the next two questions, let C_{old} be the answer to Question 1.10.

Question 1.11 (2 pts). If I double $m \mapsto 2m$, compute the new cost in terms of C_{old}.

Solution: The new cost is $2mn!$, so the new cost is $2C_{\text{old}}$.

Question 1.12 (2 pts). If I increment the number of vertices $n \mapsto n+1$, determine an expression for the new cost in terms of C_{old}.

Solution: The new cost is $m(n+1)!$, which is $(n+1)C_{\text{old}}$.

Thus, figuring out whether two graphs are isomorphic can be expensive. However, this doesn't necessarily mean that once you know the answer, *proving* to someone else that two graphs are isomorphic is expensive. Suppose there are two friends, Paula and Victor. Paula tells Victor that she thinks two graphs G_1 and G_2 are isomorphic. Victor seems skeptical, so Paula does the following to prove it.

System 1.4. For some graphs G_1 and G_2:

1. Paula hands Victor the complete evaluation table of a function, f, she claims is an isomorphism from G_1 to G_2. That is, she hands him a table with all possible inputs (the vertices of G_1) and their matching output:

vertex	$f(\text{vertex})$
v_1	w_1
v_2	w_2
$\vdots$	$\vdots$
v_n	w_n

2. Victor checks if f is actually an isomorphism. If it is, he believes Paula's claim that the graphs are isomorphic. Otherwise, he doesn't believe Paula.

For all of our systems, if Paula sends any extraneous information that is not outlined in the scheme, Victor rejects automatically (here, she HAS to send an evaluation table of the correct size, even though it may be incorrect, e.g. if we have $G_1 \not\cong G_2$).

> **Question 1.13** (2 pts). In terms of $n = |V_1|$ and $m = |E_1|$, compute the size of the evaluation table of the isomorphism f. Assume that a vertex is of size 1.
>
> **Solution**: There are n rows in the table, one for each of the n vertices. Each row has 2 columns, so the total size is $2n$.
>
> **Question 1.14** (4 pts). Write an algorithm, using the evaluation table that Paula provided, that checks Paula's claim.
>
> **Solution**: For each edge (v_i, v_j) in G_1, verify that (w_i, w_j) is in G_2. Since they have the same number of edges, we are done.
>
> **Question 1.15** (3 pts). Again, consider charging the algorithm with the cost scheme outlined in Definition 1.3. Compute the cost of your algorithm in the worst case. Any operation without an explicitly defined cost can be assumed to be free.
>
> **Solution**: \$$2m$, as we charge \$$m$ for iterating through the edge set and have to do \$$m$ lookups in G_2.

Thus, it is often much more efficient to check a proof than it is to solve a problem from scratch! This makes intuitive sense; checking is a "linear" operation, involving just checking that each step is sufficiently justified by previous steps. Presenting your own proof requires much more thinking and insight[1]. This motivates the following definition.

> **Definition 1.5.** For the purposes of a graph problem with one or more graphs G_i, an **efficient** algorithm means that the cost of the algorithm is **a polynomial** in terms of $m = \max_i |E_i|$ and $n = \max_i |V_i|$ (the exponents cannot depend on m or n). Algorithms with costs \$$2^n$ and \$$m^n$ are not efficient, but \$$m^2 n$ and \$$m^{100}$ are.

You can assume the following conjecture is true for the rest of the round (although no one has proved it yet):

> **Conjecture 1.6.** There exists no efficient algorithm in general to determine whether two graphs G_1 and G_2 are isomorphic.

We suspect that even though it's easy to check an isomorphism, it's hard to find one.

2   Probabilistically-Checkable Interactive Proofs (31 pts)

In mathematics, a proof is generally accepted if the person reading it (a verifier) finds that it is logically consistent and justifies the claims made. However, this is not the only way to prove something. For instance, if a friend claimed to know the winning numbers to a lottery ahead of time and hit the jackpot 5 times in a row, it would generally be acceptable to believe that they have a means of knowing those numbers ahead of time, even if there is a chance that they had simply been lucky in guessing. By asking them to guess more and more lottery numbers, the chances get better and better.

Suppose a newly-released, efficient algorithm is claimed to simulate fair dice rolls. In this case, the intention is that each outcome of a die has a $\frac{1}{6}$ chance of occurring. Paula claims that the algorithm is faulty, and each roll will instead always produce the same number.

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[1]This is exactly the distinction between the complexity classes $\mathcal{P}$ and $\mathcal{NP}$, the subjects of one of the Millenium Problems!

System 2.1. To prove the die-rolling algorithm returns the same result for each roll, Paula does the following.

1. Paula tells Victor what the algorithm will roll.
2. Victor uses the efficient algorithm to simulate the rolls of the die once.
3. If Victor's simulated roll of the dice matches what Paula predicted, he believes her claim that the algorithm produces the same result for each roll. Otherwise, he doesn't believe Paula since her prediction was wrong.

In this case, if the algorithm returns the same result for each roll, and Paula knows this result, she should always successfully predict the outcome of the simulated die roll. Consequently, Victor would always believe her.

Question 2.1 (2 pts)**.** Suppose the die-rolling algorithm correctly simulates a fair die, returning one of six random outcomes, each with probability $\frac{1}{6}$. In this case, Paula's claim would be false. Compute the probability that she still correctly predicts the outcome of the simulated roll, which would convince Victor that the die simulation is faulty.

Solution: If the algorithm simulates a fair die correctly, then Paula has a $\frac{1}{6}$ chance of guessing the roll correctly.

Under the right conditions, systems like the one above are **probabilistically-checkable interactive proofs**.

Definition 2.2. A **probabilistically-checkable interactive proof** (PCIP) system is a coordinated algorithm between two players, named Victor and Paula. It consists of back-and-forth communication between the two parties, wherein Paula is trying to prove a statement x to Victor, and Victor can only run *efficient* algorithms.

- The **completeness** of the system is the probability that Victor believes x is true given that Paula's claim is actually true. In other words, it measures Paula's ability to prove a true statement to Victor.
- Suppose x is false, but Paula is trying to make Victor believe it is true. The **soundness** of the system is the maximum probability, over all possible strategies of Paula (where she has to send the same messages as if she were honest), that Victor believes Paula. In other words, it measures Victor's ability to avoid believing false statements from Paula.

We require that a PCIP satisfies the following properties:

1. The completeness is at least $2/3$.
2. The soundness is at most $1/3$.

For instance, the completeness of System 2.1 is 1, while the soundness of the system is the answer to Question 2.1.

System 2.3. Suppose that the dice algorithm from System 2.1 is faulty, but returns the number 3 with probability $\frac{1}{2}$ and the other numbers each $\frac{1}{10}$ of the time. Suppose Paula knows this and makes the claim to Victor that the algorithm has these new probabilities. Paula uses the same procedure as System 2.1, where she always tells Victor that a 3 will be rolled.

Question 2.2 (2 pts). Compute the completeness of System 2.3; that is, when the distribution is indeed like this, find the probability that the system succeeds.

Solution: $\frac{1}{2}$, since that is the probability that the algorithm will roll 3.

In fact, the $1/3$ and $2/3$ values in Definition 2.2 are somewhat arbitrary. Indeed, many other sets of values work, as long as the completeness is greater than the soundness.

Definition 2.4. For a PCIP system S, define the **repetition system** $\text{REP}_{\ell,T}(S)$ as the system where the pair repeats the system S ℓ times independently (i.e. all randomness between runs is independent) with a threshold $T \in [0, 1]$, where Victor believes Paula overall if he believes her claim after at least $T\ell$ of the repetitions.

Question 2.3 (5 pts). Compute some T and ℓ such that $\text{REP}_{\ell,T}(\text{System } 2.3)$ has completeness greater than $\frac{2}{3}$ and the soundness less than $\frac{1}{3}$.

Solution: $\ell = 2$ and $T = 0.5$ are sufficient. If Paula's claim is true, then the system has a $1 - \left(\frac{1}{2}\right)^2 = \frac{3}{4} > \frac{2}{3}$ chance of succeeding at least once out of the two trials. If Paula's claim is false, then there is a $1 - (\frac{5}{6})^2 = \frac{11}{36} < \frac{1}{3}$ chance of at least one of the rolls being 3, which would falsely convince Victor of her claim.

For large enough ℓ, we can utilize the **law of large numbers**.

Theorem 2.5. The **law of large numbers** says that when a random experiment (such as a PCIP) is repeated enough times, the fraction of trials that correspond to each possible outcome gets arbitrarily close to the probability of that outcome happening.

For instance, suppose Victor has a probability p of believing Paula in a PCIP and a probability $1 - p$ of not believing her. If this PCIP is run for large enough ℓ, then Victor will believe Paula in about $p\ell$ of those trials and will not believe her in about $(1 - p)\ell$ of those trials.

Question 2.4 (5 pts). Suppose that Victor and Paula run a PCIP system S. Find an explicit threshold T where there exists an ℓ such that the soundness of $\text{REP}_{\ell,T}(S)$ is arbitrarily close to 0 and the completeness of $\text{REP}_{\ell,T}(S)$ is arbitrarily close to 1. Justify.

Solution: T can be any threshold between $\frac{1}{3}$ and $\frac{2}{3}$. Let $\epsilon > 0$. With probability $1 - \epsilon$, by law of large numbers, for large enough ℓ, if Paula's claim is true, then at least $\frac{2}{3}\ell > T\ell$ of the trials will succeed, so completeness is arbitrarily high. Similarly, if Paula's claim is false, then at most $\frac{1}{3}\ell < T\ell$ of the trials succeed, so soundness is arbitrarily low.

To illustrate another situation where a PCIP system could be useful, let's tackle a problem similar to Question 1.14. Paula and Victor still have access to two graphs G_1, G_2 with the same numbers of vertices and edges, but Paula wants to prove that the graphs are NOT isomorphic (this is her statement x).

Question 2.5 (3 pts). Explain why our previous algorithm from Question 1.14, of sending the evaluation table of an isomorphism and checking it, is insufficient for proving that two graphs are NOT isomorphic.

Solution: There are $n!$ possible evaluation tables for 2 graphs. Verifying that one of those tables is not an isomorphism does not prove that none of the other $n! - 1$ possible tables does contain an isomorphism.

Thus, we must turn to a PCIP system.

System 2.6. Consider the following system:

1. Victor selects at random one of the two graphs G_1 or G_2 and sends to Paula a random isomorphic copy of this graph, G'.

2. Upon receiving G', Paula tells Victor which of G_1 or G_2 she thinks G' was copied from (i.e. if she thinks it's G_b, then she sends the number $b \in \{1, 2\}$).

3. If Paula tells Victor the correct answer, then Victor believes that G_1 and G_2 are not isomorphic; otherwise, he rejects Paula's proof.

Question 2.6 (4 pts). State an efficient procedure to generate an isomorphic copy of a graph uniformly at random. Assume that for arbitrary N, you can generate a random number in $\{1, 2, \ldots, N\}$ for \$1. Give the cost of your procedure (still charging the set operations from before).

Solution: To generate a random isomorphic copy, we first randomly generate a function f to map the vertices of the original graph to those of its copy. To do this, we can generate random numbers from $\{1, 2, \ldots, N\}$ without replacement (so we remove a number i from the set when it is randomly generated) until all numbers have been generated. Then, we define $f(1)$ to be the first number generated, $f(2)$ to be the second number generated, and so on. Since this requires n random numbers to be generated, the cost of this is \$n$. Afterwards, the edges from the original graph need to be copied onto the copy (to ensure f is a graph isomorphism), which takes \$m$. The total cost is \$(n+m)$. **EDIT**: Since we didn't give a bound on N, you can also just set $N = n!$ and choose a permutation based off the integer in the set you pick, which will make the entire algorithm only take \$m$.

Question 2.7 (3 pts). Suppose Paula wasn't honest and the graphs were actually isomorphic. Explain why Paula has no hope, past random guessing, of figuring out which graph G' came from.

Solution: The isomorphic copy sent by Victor would be isomorphic to both graphs, so it could have been generated from either graph. Thus, Paula has no way of distinguishing between the two.

However, as is, this isn't quite a PCIP since the completeness and soundness are lacking a bit.

Question 2.8 (2 pts). Compute the completeness of this system. That is, if the graphs are not isomorphic and Paula is able to tell them apart, then compute the probability Victor believes this.

Solution: 1, since Paula doesn't need to rely on randomness to tell the graphs apart.

Question 2.9 (2 pts). Compute the soundness of this system. That is, if the graphs are isomorphic and Paula is lying, then compute the maximum probability Victor believes her. *HINT: Paula sends exactly one piece of information to Victor and cannot send anything else.*

Solution: $\frac{1}{2}$, since Paula has a $\frac{1}{2}$ chance of guessing the right graph.

Question 2.10 (3 pts). Explain how to amplify soundness and completeness to the $1/3$ and $2/3$ thresholds that are necessary, i.e. to make the resulting scheme a PCIP.

Solution: Victor can run 2 independent trials and require Paula to be correct in both trials to believe her. Completeness is still 1, but soundness is now $\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} < \frac{1}{3}$ since Paula would have to correctly guess twice.

3 Zero-Knowledge Proofs (37 pts)

System 1.4 provides a way for Paula to prove to Victor that two graphs are isomorphic. However, it requires her to give an isomorphism f to Victor. In some situations, Paula may not necessarily want to give out full information during a proof. To determine what Victor learns from running an interactive proof with Paula, we consider the **transcript of communication** of the proof.

Definition 3.1. A **transcript of communication** is a record of the messages that were exchanged between Victor and Paula.

For instance, if Victor and Paula run System 1.4, the transcript would be the entire evaluation table of f that Paula passes to Victor. Other information during this proof that wasn't shared, such as Victor's work to verify Paula's isomorphism, is not included in the transcript. In other words, the transcript of communication should only include information that is visible to both Victor and Paula during the proof.

Question 3.1 (2 pts). Describe what information is included on the transcript of communication from running System 2.6.

Solution: The transcript of communication is a graph $G' = (V', E')$ (two sets), and the number b.

Question 3.2 (4 pts). Suppose an *efficient* algorithm M exists that produces the transcript of communication for a proof system. Explain intuitively why the existence of M indicates that Victor learns no secret knowledge from interacting with Paula within the proof system. *HINT: Victor can run efficient algorithms.*

Solution: Since Victor can run M by himself, that means any interaction with Paula could be simulated and so Victor cannot gain any new information running M couldn't have given him.

Many proof systems, such as System 2.6, rely on randomness that occurs during the proof. In these cases, even when using the same graphs, the transcript of communication between Victor and Paula is not fixed due to the randomness that occurs. In this case, we consider all possible transcripts that could occur as well as the probability each one occurs. A function that describes this relation is a **probability distribution**.

> **Definition 3.2.** In general, a **probability distribution** for a random experiment is a function $\mathbb{P}$ that takes in any possible outcome as an input and outputs the probability that outcome occurs.

For example, the possible outcomes of a roll of a six-sided die are 1, 2, 3, 4, 5, and 6. Since the die is fair, for any integer $1 \le x \le 6$, $\mathbb{P}(x) = \frac{1}{6}$.

Two probability distributions are **equal in distribution** if, for each outcome in the first distribution, a corresponding outcome in the second distribution also has the same probability. For instance, the distribution of whether the roll of a fair six-sided die is even or odd and the distribution of the result of flipping a fair coin are equal, since the probability of each event (even or odd, heads or tails) is $\frac{1}{2}$.

> **Question 3.3** (3 pts)**.** Give another example of two simple random experiments whose outcomes are equal in distribution, but the outcomes are not necessarily the same.
>
> **Solution**: Another example could be spinning a spinner with 10 equal sections, where 2 of them are winning and spinning another with 10 equal sections, where 8 of them are winning. Then, the distributions of winning/losing in both are equal, though the outcomes which are 20% are winning on the first spinner and losing on the second spinner, which are different outcomes per se.

Since a transcript of communication contains all messages Victor receives from Paula during a proof, we can use this transcript to determine whether or not Victor learned any unnecessary information during the proof. Proofs where Victor does not learn any additional information are said to be **zero-knowledge**.

> **Definition 3.3.** A **zero-knowledge proof** (ZKP) system to prove a statement x is a PCIP system and an *efficient* algorithm M (called the **simulator**) where the output generated by M on input x is equal in distribution to Paula and Victor's transcript of communication (in the case where the statement is correct and Paula knows the proof). That is, if T is some transcript of messages, we must have $\mathbb{P}(M(x) = T) = \mathbb{P}(\text{Paula and Victor make } T)$.

In a sense, M "indistinguishably simulates" a possible communication between Victor and Paula.[2]
We can reanalyze our previous systems and see if they have the zero-knowledge property.

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[2]Technically, this is the notion of honest-verifier perfect zero-knowledge, but that distinction does not matter for us here.

Question 3.4 (6 pts). Show that the algorithm you designed in Question 1.14 is a PCIP system, but not a ZKP system. *HINT: You may assume conjecture 1.6, that there is no efficient way to tell if two graphs are isomorphic.*

Solution: Note that if Paula has the evaluation table, then with probability 1 all the edges will be satisfied and Victor will be convinced, so completeness is 1. For soundness, if the graphs are not isomorphic, then there exists no isomorphism between them; therefore, no matter what evaluation table Paula sends, Victor will not believe her. Thus, this is a PCIP system.
However, this is not a ZKP system. If it was, then there would exist an efficient simulator M which would just make the evaluation table (which is not random, so the distribution is just the isomorphism with probability 1), i.e. it would find the isomorphism. This would efficiently solve Graph Isomorphism, which is impossible under conjecture 1.6.

Question 3.5 (4 pts). Show that System 2.6 is a ZKP system.

Solution: We have already shown it's a PCIP system, so it's sufficient to show the zero-knowledge property. To create a simulator, we can select a graph ($1, create a random isomorphic copy (answer to question 2.6), then return the correct selection from Paula's side (free) all efficiently. Clearly their sum is still polynomial, so we're done.

Now, let's refine the scheme from Question 1.14 to make it zero-knowledge. We will do this by introducing some randomness. Suppose we have two graphs G_1, G_2 that Paula wants to prove are isomorphic.

System 3.4. Consider the following system, where Paula claims she knows an isomorphism f from G_1 to G_2. Assume that $|V_1| = |V_2|$.

1. Paula chooses a random bijection g and sends $H = g(G_2)$, the graph you get by putting G_2 through this isomorphism.
2. Victor chooses a random number $b \in \{1, 2\}$ and sends b to Paula.
3. Paula then sends the evaluation table of a bijection h from the vertices of G_b to the vertices of H. If $b = 1$, $h(v) = g(f(v))$. If $b = 2$, $h(v) = g(v)$.
4. Victor believes Paula if h is an isomorphism (it respects edges).

Let's analyze this scheme.

Question 3.6 (4 pts). Compute the soundness of this scheme.

Solution: Paula's best strategy is as follows. She chooses some isomorphism g; it doesn't matter what it is. If $b = 2$, she can return back the isomorphism, and if $b = 1$ she cannot return $g \circ f$ because since G_2 and H are isomorphic and G_1 and G_2 are not, it follows G_1 and H are not. Thus, she can succeed with probability at most $\frac{1}{2}$.

Question 3.7 (2 pts). Compute the completeness of this scheme.

Solution: We again have completeness 1, as no matter what, Paula will present a correct h if she has f, if $b = 1$, $g(f(v))$ is still an isomorphism (and in the $b = 2$ case she will trivially give an isomorphism regardless of whether she knows f).

Question 3.8 (8 pts). Note that the transcript of messages sent is the triple $(H, b, \text{evaluation table of } h)$. Find an efficient algorithm M to generate the transcript between the two players. Analyze the cost of your algorithm to show it is efficient.

Solution: Do these things out of order. First, sample a random bit $b \in \{1, 2\}$. Now, make a random isomorphism h and define $H = h(G_b)$. We note that

1. If $b = 1$, $H = h(G_1) = h(f^{-1}(G_2))$, so $g = h \circ f^{-1}$ is a uniformly random isomorphism between the two graphs. To see this, note that f^{-1} is one-to-one, so for all the x's, $f^{-1}(x)$ gets mapped to different outputs. Further, h is equally likely to map any input y to any output z, so each $f^{-1}(x)$ is equally likely to be mapped to any output and each x gets mapped to a distinct output, making it a random isomorphism. We ought to return $g \circ f = h \circ (f^{-1} \circ f) = h$, so we are returning the correct function.

2. If $b = 2$, $H = h(G_2)$ so $g = h$ is a random isomorphism from G_2 to H and thus we return the correct function.

As before, sampling a random bit costs \$1, and sampling a random isomorphism costs \$$n$, and iterating through E_2 to use the look-up table costs \$$m$. Overall, this is still polynomial in m and n. Thus, using the completeness/soundness amplification we noted prior, System 3.4 is a ZKP system.

Now that we've found examples and non-examples of ZKP systems, let's consider how they behave when used together. For instance, suppose Victor wanted to solve question Question 1.5 with help from Paula. System 2.6 only allows Victor to ask about 2 graphs at a time, but suppose he wants to check 8. To resolve this, he could just run the system multiple times, once with each combination of 2 different graphs.

Definition 3.5. Given ℓ PCIP Systems $S_1, S_2, \ldots, S_\ell$, their **serial composition** is the result of running them one after another independently. That is, Paula tries to convince Victor a statement x_1 is true through a PCIP protocol S_1, then convinces Victor about a (possibly different) statement x_2 through S_2, and so on. All messages related to the system S_j must be sent/received before the first message of S_k, for all $j < k$. You may assume that ℓ is independent of the time it takes to run each PCIP system.

Question 3.9 (4 pts). Show that the serial composition of multiple zero-knowledge proofs will always result in an interaction with the zero-knowledge property.

Solution: Because each proof in the serial composition is independent of the others and has the zero-knowledge property, a simulator M_i exists for each S_i. A transcript for the serial composition consists of the transcripts produced all put together from each individual proof. This transcript can be generated efficiently by running each of the simulators $M_1, M_2, \ldots, M_\ell$ for the individual proofs sequentially, so an algorithm that runs the simulators for the individual proofs in order is a simulator for the concatenation of proofs. Since ℓ is a constant and the constant sum of polynomials is still polynomials, this simulator is efficient.

4 ZKP Systems From Other Hardness (32 pts)

In computer science, we often struggle to find efficient algorithms for problems, so we conjecture that they are hard. As we saw with graph isomorphisms, assuming this allows us to get zero-knowledge schemes. Let's use another common conjecture to form another zero-knowledge proof scheme. For the purposes of this section, p is a very large prime number.

Definition 4.1. An integer $g \pmod{p}$ is called a **generator** if every number in $\{1, 2, \ldots, p-1\}$ can be written as $g^a \pmod{p}$ for some a.

Here, in the modular arithmetic setting, we charge costs a little differently: adding or multiplying two numbers $\bmod\ p$ costs \$1. Making random numbers still costs the same. An algorithm is **efficient** in modular arithmetic if it's a polynomial in $\log_2 p$, the number of binary digits in p.

Question 4.1 (6 pts). Devise an efficient algorithm for computing $g^a \bmod p$.

Solution: One can use the following algorithm. Write a in its binary representation $a = a_{\lceil \log_2 p \rceil} a_{\log_2 p - 1} \ldots a_1 a_0$. Then, we can compute $g, g^2, g^4, \ldots, g^{2^k}$ for $2^k > p$ in $\lceil \log_2 p \rceil$ multiplications by repeatedly squaring. Then we compute $g^a = \prod_i a_i g^{2^i} \bmod p$. This is at most $\lceil \log_2 p \rceil$ multiplications mod p, so we end up with a cost less than $3 \log_2 p$, which is efficient.

It turns out that undoing the operation is much more difficult.

Conjecture 4.2 (Discrete Logarithm Assumption). Given a generator g and $g^a \bmod p$ for some a in $\{1, 2, \ldots, p-1\}$ that you do not know, there is no expected efficient algorithm to find a (i.e. one whose cost is, on average over all randomness, polynomial in $\log_2 p$).

Before we can harness this, we should see why modular arithmetic plays nicely with randomness.

Question 4.2 (3 pts)**.** Show that given a fixed number N, then if we randomly pick R in $\{0, 1, \ldots, p-1\}$, then $N + R \bmod p$ is equal in distribution to R.

Solution: Note that for a fixed $r \in \{0, 1, \ldots, p-1\}$, $\mathbb{P}(N + R \equiv r \pmod{p}) = \mathbb{P}(R \equiv r - N \pmod{p}) = \frac{1}{p}$, giving all the outcomes the same probabilities as $\mathbb{P}(R = r) = \frac{1}{p}$.

Question 4.3 (3 pts)**.** Show that given a fixed number N not equivalent to $0 \pmod{p}$, if we randomly pick R in $\{1, 2, \ldots, p-1\}$, then $NR \bmod p$ is equal in distribution to R.

Solution: Note that for a fixed $r \in \{1, \ldots, p-1\}$, $\mathbb{P}(NR \equiv r \pmod{p}) = \mathbb{P}(R \equiv N^{-1}r \pmod{p}) = \frac{1}{p-1}$, giving all the outcomes the same probabilities as $\mathbb{P}(R = r) = \frac{1}{p-1}$.

Suppose now that everyone has access to the same prime p and a generator $g \pmod{p}$. Paula picks a random number α from the set $\{1, 2, \ldots, p-1\}$ and gives Victor $u = g^\alpha$. Victor and Paula want to make a scheme so Victor can identify Paula in communications, without Victor himself being able to impersonate Paula.

Question 4.4 (20 pts)**.** Construct a ZKP scheme that allows Paula to convince Victor that she knows α, and does not allow others to convince Victor they know α. Here is a possible scheme with some steps removed that you can use as a template.

1. When she wants to log in, Paula chooses randomly $\alpha_t \in \mathbb{Z}_p$ (where $\mathbb{Z}_p$ is the set $\{0, 1, 2, \ldots, p-1\}$), computes $u_t =$ _____ $\bmod p$ and sends u_t to Victor.
2. Victor chooses randomly $c \in \mathbb{Z}_p$ and sends c to Paula.
3. Paula computes $\alpha_z =$ _____ $\bmod p$ and sends it to Victor.
4. Victor accepts the proof if $g^{\alpha_z} \equiv$ _____ $\pmod{p}$.

Discuss the soundness and completeness of your scheme, and provide an efficient simulator for the transcript.

Solution: **EDIT:** We forgot to specify that "other" people who attempt to log in are computationally bounded, i.e. they can only run efficient algorithms. If an adversary is all-powerful, they can just brute force to find α such that $u = g^\alpha$.
One solution is as follows.
Protocol

1. When she wants to log in, Paula chooses randomly $\alpha_t \in \mathbb{Z}_p$, computes $u_t = g^{\alpha_t} \bmod p$ and sends u_t to Victor.

2. Victor chooses randomly $c \in \mathbb{Z}_p$ and sends c to Paula.

3. Paula computes $\alpha_z = \alpha_t + \alpha c \bmod p$ and sends it to Victor.

4. Victor accepts the proof if $g^{\alpha_z} \equiv u_t \cdot u^c \pmod{p}$.

Completeness Now, if Paula has α, then this equation will always be true as

$$g^{\alpha_z} \equiv g^{\alpha_t + \alpha c} \equiv g^{\alpha_t} \cdot (g^\alpha)^c \equiv u_t \cdot u^c \pmod{p}$$

and the completeness of the scheme is 1.
Soundness Suppose that the prover is not Paula but instead Patty, who doesn't know α. Assume for the sake of contradiction, Patty had an efficiently computable strategy (which costs some polynomial $P(\log_2 p)$) with probability $s \geq 1/3$ of succeeding a trial. Then, she has some adversarial strategy for

choosing u_t. Then at this point, if Victor chooses some random c in step 2, Paula has a follow-up α_z. The probability that this (c, α_z) pair works is s, by definition. Furthermore, if Victor chooses some other random c' in step 2, Patty similarly has a follow-up α'_z where this is accepted with probability s again. With probability $1 - \frac{1}{p}$, we have that $c \neq c'$. Suppose someone ran this scheme as an algorithm in a black box, playing both the role of Patty and Victor. Since it doesn't require α, we claim this violates the Discrete Logarithm assumption for general u, i.e. one can find α for arbitrary u. Since Victor believed Patty was Paula in both cases, we must have that $g^{\alpha_z} \equiv u_t \cdot u^c \pmod{p}$ and $g^{\alpha'_z} \equiv u_t \cdot u^{c'} \pmod{p}$. This yields

$$u_t \equiv g^{\alpha_z} u^{-c} \equiv g^{\alpha'_z} u^{-c'} \pmod{p}$$
$$g^{\alpha_z - \alpha'_z} \equiv u^{c-c'}$$
$$(g^{\alpha_z - \alpha'_z})^{1/(c-c')} \equiv u \equiv g^\alpha$$

Thus, we can calculate that $\alpha = \frac{\alpha_z - \alpha'_z}{c - c'} \pmod{p-1}$ for any u. We will finally show that this algorithm would be efficient. In expectation, it takes $\frac{p}{(p-1)s^2} < 2 \cdot 9 < 20$ repetitions to get $c \neq c'$ such that Victor accepts Patty's proof in both cases. The algorithm thus has expected cost $20 \times$ cost of one run and the cost of one run can be seen to be at most $\$(K_1 + 3K_2 \log_2 p + P(\log_2 p)$ for some constants K_1, K_2, as we do a constant amount of multiplications/additions and three exponentiations. We thus have an expected efficient algorithm that violates Conjecture 4.2. Thus, soundness is less than $1/3$.

Efficient Simulator Finally the transcript is (u_t, c, α_z). c can be sampled efficiently for \$1. Now, α_z looks random by the previous two problems, even though c is now fixed. Thus it is sufficient to just sample $\alpha_z \in \mathbb{Z}_p$ at random with another dollar. Computing $u_t = g^{\alpha_z} \cdot u^{-c}$ requires at most 3\$$\log_2 p$. Thus, the cost of the algorithm is at most, \$$3 \log_2 p + 3$, which is a polynomial in $\log_2 p$. Thus, the system is indeed a ZKP.