## Question 1

One method for gauging how hard it is to prove a conjecture $C$ is to investigate if not $C$ is true under the assumption that P equals NP. If $\mathrm{P}=\mathrm{NP}$ implies not $C$ then proving $C$ would also prove that $\mathrm{P} \neq \mathrm{NP}$, so a proof of $C$ (if true) is likely out of reach. Show that the following statements are true assuming $\mathrm{P}=\mathrm{NP}$ :
(a) The problem "Given a circuit $C$ as input, all assignments are satisfying for $C$ " is in P .

Solution: On input $C$, apply the polynomial-time algorithm for SAT to the circuit not $C$ and negate the answer. If all assignments to $C$ are satisfying then Not $C$ has no satisfying assignment and the procedure accepts. Otherwise nот $C$ has a satisfying assignment and the procedure rejects.
(b) Polynomial Identity Testing is in P. (Hint: Think of the randomness as a potential NP certificate.)

Solution: In Lecture 7 we showed that for every size- $s$ instance $C$ of polynomial identity testing that does not compute the identically zero polynomial a random assignment of the inputs from $\{1, \ldots, 3 s\}$ evaluates to zero with probability at most $1 / 3$. In particular every nonzero $C$ has at least one polynomial-size witness $x \in\{1, \ldots, 3 s\}^{n}$ such that $C(x)$ does not evaluate to zero. Therefore the set of pairs $(C, x)$ where $C(x) \neq 0$ and $x \in\{1, \ldots, 3 s\}^{n}$ is an NP-relation whose decision version is the complement $\overline{P I T}$ of polynomial identity testing. If P equals NP then $\overline{P I T}$ is in P and so is PIT itself as P is closed under complement.
(c) There is no polynomial-time computable family $G_{n}:\{0,1\}^{n} \rightarrow\{0,1\}^{n+1}$ of $\left(2^{n / 10}, 1 / 4\right)$-pseudorandom generators. (Hint: The problem "On input $y$, does there exists $x$ such that $G_{|x|}(x)=y$ ?" is in NP.)

Solution: Every output of $G_{n}$ is a YES-instance of the problem described in the hint. The probability that a random string $Z$ in $\{0,1\}^{n+1}$ is a yes instance is at most half because only half of the strings are possible outputs of $G_{n}$. If P equals NP the polynomial-time algorithm $D$ for this problem is a distinguisher with advantage at least $1-1 / 2>1 / 4$. In particular this algorithm can be implemented by a family of polynomial-size circuits which fits within the required bound of $2^{n / 10}$ for all sufficiently large $n$.

## Question 2

Assume $f:\{0,1\}^{n} \rightarrow\{0,1\}$ is 0.01 -unpredictable against size $n^{2}$. Which of these constructions is an $\left(n^{2} / 10,0.1\right)$ pseudorandom generator? If you answer no describe a distinguisher for $G$. If you answer yes show how to convert a distinguisher for $G$ into a predictor for $f$ (possibly using results from class). Addition denotes (bitwise) xor.
(a) $G:\{0,1\}^{2 n} \rightarrow\{0,1\}^{2 n+2}$ given by $G(x, y)=(x, f(x), y, f(x)+f(y))$.

Solution: Yes. If not suppose $D$ has size $n^{2} / 10$ and 0.1 -distinguishes $(x, f(x), y, f(x)+f(y))$ from a random string. Let $D^{\prime}$ be the circuit that takes input $(x, a, y, b)$ and applies $D$ to $(x, a, y, a+b)$. Then $D^{\prime}$ has size $n^{2} / 10+O(1)$ and 0.1 -distinguishes $(x, f(x), y, f(y))$ from a random string $(x, a, y, b) . D^{\prime}$ must then 0.05 -distinguish either of those from $(x, f(x), y, b)$. In one case by fixing $x$ (and $f(x)$ ) that maximizes the advantage of $D^{\prime}$ we get a 0.05 -distinguisher of $(y, f(y))$ from a random string of size $n^{2} / 10+O(1)$. In the other case by fixing $y$ and $b$ in an advantage-maximizing way we get a distinguisher of ( $x, f(x)$ ) from random of the same advantage and size. In either case by Yao's lemma $f$ can be 0.05 -predicted by size $n^{2} / 5+O(1)$ contradicting the assumption.
(b) $G:\{0,1\}^{n m} \rightarrow\{0,1\}\binom{m}{2}$ (one output for every pair of inputs), with $m=3 n$, given by

$$
G\left(x_{1}, \ldots, x_{m}\right)=\left(f\left(x_{1}\right)+f\left(x_{2}\right), \ldots, f\left(x_{1}\right)+f\left(x_{n}\right), f\left(x_{2}\right)+f\left(x_{3}\right), \ldots, f\left(x_{m-1}\right)+f\left(x_{m}\right)\right)
$$

Solution: No. The output of $G$ includes the three bits $f\left(x_{1}\right)+f\left(x_{2}\right), f\left(x_{1}\right)+f\left(x_{3}\right)$, and $f\left(x_{2}\right)+f\left(x_{3}\right)$ which always XOR to zero. The distinguisher that computes the XOR of these three bits always accepts
outputs of $G$ but only accepts random strings with probability half, showing that $G$ is not even $(O(1), 1 / 2)$ pseudorandom.
(c) (Optional) $G:\{0,1\}^{3 n} \rightarrow\{0,1\}^{3 n+3}$ given by $G(x, y, z)=(x, y, z, f(x+y), f(x+z), f(y+z))$.

Solution: I don't know the answer to this one.

## Question 3

In Lecture 3 we showed that the following property of functions $f:\{0,1\}^{n} \times\{0,1\}^{m} \rightarrow\{0,1\}$ separates EQU ALITY (when $m=n$ ) from width $2^{n}$ read-once branching programs:
$\operatorname{diffext}(f)$ : For every pair $x \neq x^{\prime} \in\{0,1\}^{n}$ there exists a $y \in\{0,1\}^{m}$ such that $f(x, y) \neq f\left(x^{\prime}, y\right)$.
(a) Argue that diffext is $2^{O(n+m)}$-constructive, namely describe an efficient algorithm that decides diffext $(f)$ using oracle access to $f$ and analyze its running time.

Solution: The algorithm loops over all $\binom{2^{n}}{2}$ pairs $x \neq x^{\prime}$. For each of these pairs it tests whether any $y \in\{0,1\}^{m}$ violates the condition $f(x, y) \neq f\left(x^{\prime}, y\right)$. This condition can be checked in time $O(n+m)$, so the whole algorithm can be implemented in time $O\left((n+m) 2^{2 n+m}\right)$. This is at most quadratic in the instance size $2^{n+m}$.
(b) Show that the probability that $\operatorname{diffext}(R)$ holds for a random function $R$ is at least $1-2^{2 n-2^{m}-1}$. (Hint: Calculate the probability $R(x, y)=R\left(x^{\prime}, y\right)$ for fixed $x \neq x^{\prime}$ and all $y$ and take a union bound.)

Solution: For fixed $x \neq x^{\prime}$ the $2^{2 m}$ values $R(x, y)$ and $R\left(x^{\prime}, y\right)$ as $y$ ranges over $\{0,1\}^{m}$ are uniform and independent, so the $2^{m}$ events $R(x, y)=R\left(x^{\prime}, y\right)$ are independent of probability $1 / 2$ each. Therefore the probability that $R(x, y)=R\left(x^{\prime}, y\right)$ for all $y$ is exactly $2^{2^{m}}$. By a union bound the probability that there exist $x \neq x^{\prime}$ for which this is the case is at most $\binom{2^{n}}{2} 2^{2^{m}} \leq 2^{2 n-1} \cdot 2^{2^{m}}$.
(c) Use part (b) to show that $\operatorname{diffext}(f)$ is $1 / 2$-large (and therefore natural) when $m \geq \log (2 n)$.

Solution: When $m \geq \log (2 n)$ the probability in part (b) is at least $1-1 / 2=1 / 2$ so diffext is $1 / 2$-large as desired.

