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Can we relax the security definition in a meaningful way?

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Do we need to be concerned?

## Computational secrecy

We relax perfect secrecy in two ways:

- We allow secrecy to fail with some tiny probability
- We only restrict our attention to "efficient" attackers


## Computational secrecy

Our starting point is the following (equivalent) definition of perfect secrecy:

Definition: A private key encryption scheme $\Pi=(G e n, E n c, D e c)$ with message space $\mathcal{M}$ is perfectly indistinguishable if for every $\mathcal{A}$ it holds:

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Two possible approaches:

- Concrete
- Asymptotic


## Reminder: Perfect indistinguishability



## Computational secrecy (concrete)

Candidate definition: A private key encryption scheme $\Pi=(G e n, E n c, D e c)$ is $(t, \varepsilon)$-indistinguishable if for every attacker $\mathcal{A}$ running in time at most $t$, it holds that:

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Observation: $(\infty, 0)$-indistinguishability is equivalent to perfect indistinguishability

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## Does not lead to a clean theory

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Measure probabilities and running times as a function of $n$

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As a special case, the product of two negligible functions is negligible

## Negligible and polynomially bounded functions

Which of the following functions are polynomially bounded? Which are negligible?

$$
\begin{array}{llll}
n^{2}+4 n-2 & n^{100} & n^{3}+\cos (n) & n! \\
\frac{1}{n^{10}}+2^{-n / 2} & 2^{n} & 3^{-n} & \sqrt[3]{n}+\frac{1}{n} \\
n^{-n} \cdot\left(n^{5}+n^{2}\right) & 2^{\sqrt{n}} & \sqrt{n} & 42-\frac{1}{1+\log n} \\
4^{\sqrt{\log n}} & n^{-5} & 2^{-\log n \cdot \log \log n} & \left(1+\frac{1}{n}\right)^{n}
\end{array}
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$$
n^{-n} \cdot\left(n^{5}+n^{2}\right)
$$

$4^{\sqrt{\log n}}$

$$
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$$
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$$

$42-\frac{1}{1+\log n}$
$\left(1+\frac{1}{n}\right)^{n}$

## Negligible and polynomially bounded functions

Which of the following functions are polynomially bounded? Which are negligible?

$$
\begin{array}{llll}
n^{2}+4 n-2 & n^{100} & n^{3}+\cos (n) & n! \\
\frac{1}{n^{10}}+2^{-n / 2} & 2^{n} & 3^{-n} & \sqrt[3]{n}+\frac{1}{n} \\
n^{-n} \cdot\left(n^{5}+n^{2}\right) & 2^{\sqrt{n}} & \sqrt{n} & 42-\frac{1}{1+\log n} \\
4^{\sqrt{\log n}} & n^{-5} & 2^{-\log n \cdot \log \log n} & \left(1+\frac{1}{n}\right)^{n}
\end{array}
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- Dec is a deterministic polynomial-time algorithm that takes as input a key $k \in \mathcal{K}$ and a ciphertext $c \in \mathcal{C}$ and outputs a message $m \in \mathcal{M}$ or an error, denoted by $\perp$, if $c$ cannot be obtained by encrypting $m$.


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If $M=\{0,1\}^{\ell(n)}$ then (Gen, Enc, Dec) is a fixed-length private-key encryption scheme
(for messages of length $\ell(n)$ )

## The adversarial indistinguishability experiment, revisited



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## probabilistic polynomial-time algorithm with input $1^{n}$



路

$b \leftarrow\{0,1\}$

if $b^{\prime}=b$
if $b^{\prime} \neq b$

## The adversarial indistinguishability experiment, revisited



## The adversarial indistinguishability experiment, revisited

Adversary $\mathcal{A}$

> probabilistic polynomial-time
> algorithm with input $1^{n}$

Verifier

challenge ciphertext $\quad c \leftarrow \operatorname{Enc}_{k}\left(m_{b}\right)$


$$
\text { if } b^{\prime}=b
$$

Notation includes the security parameter

$$
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## Computational indistinguishability (asymptotic)

Definition: A private key encryption scheme $\Pi=(G e n, E n c, D e c)$ has indistinguishable encryptions in the presence of an eavesdropper (is EAV-secure) if, for every probabilistic polynomial-time adversary $\mathcal{A}$, there is a negligible function $\varepsilon$ such that:

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Observation: perfect indistinguishability implies EAV-security

## Example 1

Consider a scheme where:

- Gen $\left(1^{n}\right)$ returns a key chosen uniformly at random in $\{0,1\}^{n}$
- The best possible adversary $\mathcal{A}$ performs a brute-force search over the key space
- If the running time of the adversary is $t(n)$ then:

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Is this scheme EAV-secure? Yes!

For all polynomial running times $t(n)$, all functions in $O\left(\frac{t(n)}{2^{n}}\right)$ are negligible

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Consider a scheme where:

- $\operatorname{Enc}_{k}(m)$ runs in $n^{2} \cdot|m|$ steps
- Breaking the scheme requires $2^{n}$ steps


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How large do we need to choose $n$ ?

| $n$ | 48 | 64 | 128 | 256 | 512 | 1024 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| running time | 2.5 months | 6 months | 4 years | 32 years | 255 years | 2041 years |
| probability of success | 1 in 256 | $\approx 1$ in 17 mil | $\approx 3$ in $10^{26}$ | $\approx 3$ in $10^{65}$ | $\approx 1$ in $10^{142}$ | $\approx 2$ in $10^{296}$ |

## Computational secrecy (asymptotic)

We relax perfect secrecy in two ways:

- We allow secrecy to fail with some tiny probability $\longleftarrow$ probabilities that are negligible in $n$
- We only restrict our attention to "efficient" attackers $\qquad$ polynomial running times

Are both relaxations needed?

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Not negligible!


## Leaking the length of the message

In general, encryption does not hide the plaintext length

- This is captured in the indistinguishably experiment by requiring $\left|m_{0}\right|=\left|m_{1}\right|$


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In Google maps, the map tiles are compressed and (essentially) static. The size of the ciphertext can be used to determine the viewed location

## Where do we stand?

- We have a perfectly secret encryption scheme (one-time pad)...
- ... but it requires long keys
- This is inevitable if we insist on perfect secrecy (recall that, in a perfectly secret scheme, $|\mathcal{K}| \geq|\mathcal{M}|$ )


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It depends...

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It is widely believed that $P \neq N P$, although this would not imply that PRGs exist...

Pragmatic approach: assume that PRGs exist (and hope for the best)

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Informally, we sometimes say that $x$ is "random / uniform" to mean that it was sampled from a random/uniform distribution...
$\ldots$ and that $x$ is "pseudorandom" if it is the output of a PRG

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We would like a PRG to pass all conceivable statistical tests!

## Pseudorandomness

Historically, a candidate PRG was considered good if its outputs were able to pass a collection of statistical tests (that would be satisfied by "truly random" strings)

## Examples:

- Is the first bit of the output 1 with probability $\approx \frac{1}{2}$ ?
- Is the parity of any subset of bits 1 with probability $\approx \frac{1}{2}$ ?
- If I interpret the string as a series of points in a square of side 2 centered in the origin, is the fraction of points within the circle of radius 1 centered in the origin $\approx \pi / 4$ ?


What if somebody comes up with a new, clever statistical test we did not think of before?
We would like a PRG to pass all conceivable statistical tests!
Is this even possible?

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- At least half of the $\ell$-bit strings (actually a $\frac{2^{\ell-n}-1}{2^{\ell-n}}$-fraction) can never be output by $G$ !



## Pseudorandomness

$G$ will never pass the following statistical test (for some $n$ ):

- Look at a "sufficiently many" output/random strings
- If there are more than $2^{n}$ distinct strings, the test is passed
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Observation: This is not an efficient test.
Idea: If adversaries are polynomially bounded, we only need to pass statistical tests that run in polynomial time


## Pseudorandom Number Generators (formal)

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We need to come up with a distinguisher $D(w)$ that guesses whether $w$ comes from the output of $G(s)$ or it is chosen u.a.r. from $\{0,1\}^{\ell(n)}$

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$\left|1-\frac{1}{2^{\ell(n)}}\right|$ is not negligible


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As far as polynomial-time algorithms are concerned, the output of $G(s)$ with a random seed $s$ is indistinguishable (up to some negligible probability) from a random string $r$


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If we have a randomized polynomial-time algorithm that uses $\ell(n)$ random bits, and we replace those random bits with the output of $G(s)$, the resulting (randomized) algorithm
"behaves the same" except for a negligible probability

