

Jean Goubault-Larrecq

Randomized complexity classes

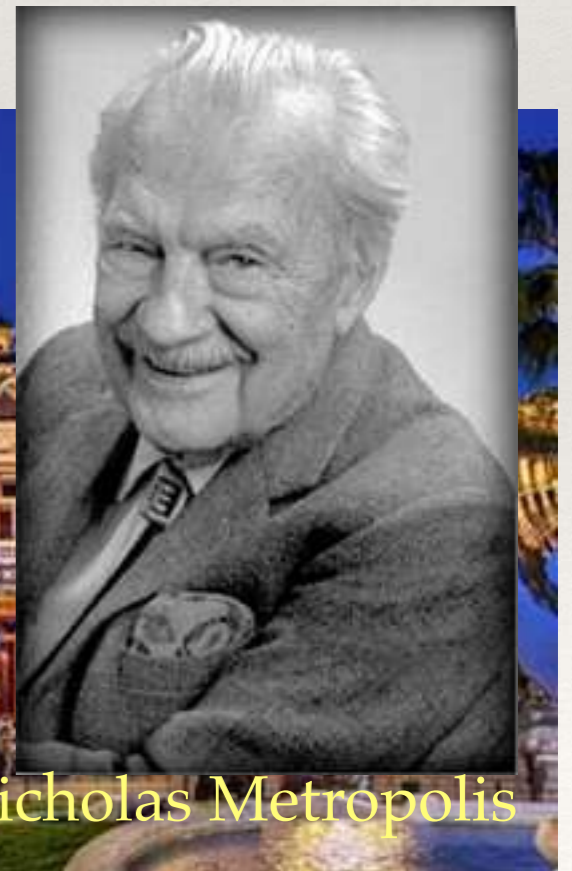
Today: **BPP** (part 1)

Today

- ❖ Two-sided error: **BPP**
- ❖ Error reduction, voting, Chernoff's bound
- ❖ The Sipser-Gács-Lautemann theorem

Our third probabilistic class: BPP

(also sometimes known as the class of *Metropolis* languages, although some speak of Monte Carlo here again)



Nicholas Metropolis

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two-sided error:
 $\Pr_r [\mathcal{M}(x,r) \text{ errs}] \leq 1/3$

Examples



<https://compeap.com/wp-content/uploads/Land-of-I-Dont-Know.jpg>

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PolyMath

Examples

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The complexity class BPP

The problem of determining whether a [multivariate polynomial vanishes](#) is in BPP. The idea of the randomized algorithm is to compute the polynomial at a small number of randomly chosen points. For a non-zero polynomial the probability that it vanishes at all those points decreases rapidly with the number of points, and so if it vanishes at all those points we can say with some confidence that the polynomial vanishes everywhere. This problem is also in co-RP, since if the polynomial really does vanish everywhere, then the algorithm is guaranteed to output 1.

It would be good to have more examples. In particular, it would be nice to have an example that isn't obviously in RP or co-RP.

https://asone.ai/polymath/index.php?title=The_complexity_class_BPP

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Error reduction

error = 1/3 here

- ❖ What is so special about error 1/3?

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- if $x \in L$ then $\Pr_r [\mathcal{M}(x,r) \text{ accepts}] \geq 1-\epsilon$
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- ❖ What is so special about error $1/3$?
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❖ **Theorem.** $\forall \varepsilon \in]0, 1/2[$,
 $\text{BPP} = \text{BPP}(\varepsilon)$.

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error = ε

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- ❖ What is so special about error 1/3?
- ❖ Nothing!

❖ **Theorem.** $\forall \varepsilon \in]0, 1/2[$,
BPP = BPP(ε).

- ❖ Note: **BPP = BPP(1/3)** (def.)
BPP(ε) = {all languages} if $\varepsilon \geq 1/2 \dots$
BPP(0) = P

error = 1/3 here

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error = ε

The easy cases: error amplification(!)

- ❖ Clearly, if $\eta \leq \varepsilon$ then
$$\mathbf{BPP}(\eta) \subseteq \mathbf{BPP}(\varepsilon)$$
- ❖ Note: $\mathbf{BPP}(0)=\mathbf{P}$ (sometimes believed $\neq \mathbf{BPP}$)
$$\mathbf{BPP}(\varepsilon)=\{\text{all languages}\}$$
 for every $\varepsilon \geq 1/2$
- ❖ In the middle, hence, we will see that all the intermediate $\mathbf{BPP}(\varepsilon)$ ($\varepsilon \in]0, 1/2[$) are equal to \mathbf{BPP} .

Error reduction

- ❖ We will show that **BPP** (= **BPP**(1/3)) is included in **BPP**(ε) for every $\varepsilon \in]0, 1/2[$, arbitrarily close to 0.

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- ❖ The technique we used for **RP** does **not** work: why?

The hard direction: repeating experiments

- ❖ Let $L \in \mathbf{RP}(\epsilon)$, $0 < \epsilon < 1$
 - ❖ On input x , let us do the following (at most) K times:
 - ❖ Draw r at random, simulate $\mathcal{M}(x, r)$ and:
 - ❖ If $\mathcal{M}(x, r)$ accepts, then exit the loop and **accept**;
 - ❖ Otherwise, proceed and loop.
 - ❖ At the end of the loop, **reject**.
- error = ϵ
- Remember: if $\mathcal{M}(x, r)$ accepts, then x **must** be in L .

Error reduction

- ❖ We will show that **BPP** (= **BPP**(1/3)) is included in **BPP**(ϵ) for every $\epsilon \in]0, 1/2[$, arbitrarily close to 0.
- ❖ The technique we used for **RP** does **not** work: why?
- ❖ Hence we must proceed differently

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❖ A language L is in **RP**(ϵ) and only if there is a **polynomial-time** TM \mathcal{M} such that for every input x (of size n):

- ❖ if $x \in L$ then $\Pr_r [\mathcal{M}(x, r) \text{ accepts}] \geq 1 - \epsilon$
- ❖ if $x \notin L$ then $\mathcal{M}(x, r)$ accepts for no r (i.e., $\Pr_r [\mathcal{M}(x, r) \text{ accepts}] = 0$).
error = ϵ

Remember: if $\mathcal{M}(x, r)$ accepts, then x **must** be in L .

Majority voting

- ❖ Imagine running $\mathcal{M}(x,r)$ for various values of r , and **tallying the votes**

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- ❖ Here 3 accepts / 1 reject
 \Rightarrow majority is for **acceptance**

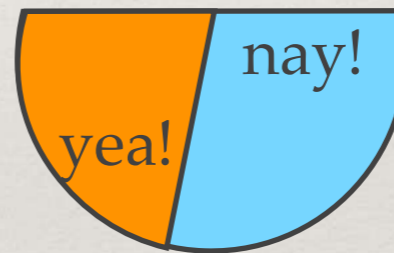


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running a large number of votes should produce a majority of **accepts**, with **high probability**

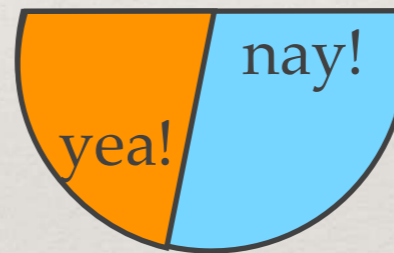


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- ❖ ... but how high?

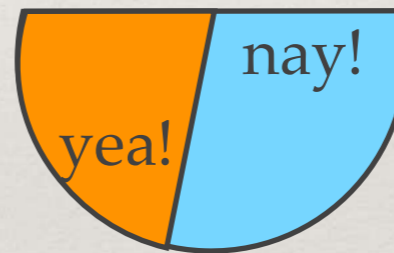


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Chernoff's bound

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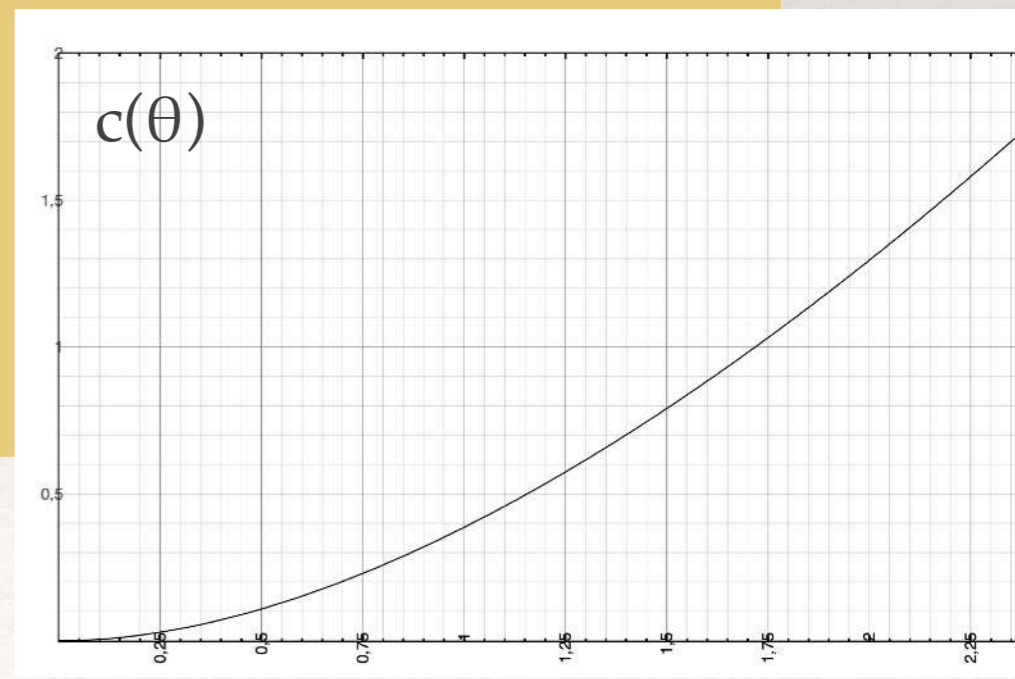
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Then $\Pr(\text{proportion of yeses among } N \text{ voters is close to } p)$
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with values in $\{0, 1\}$ and
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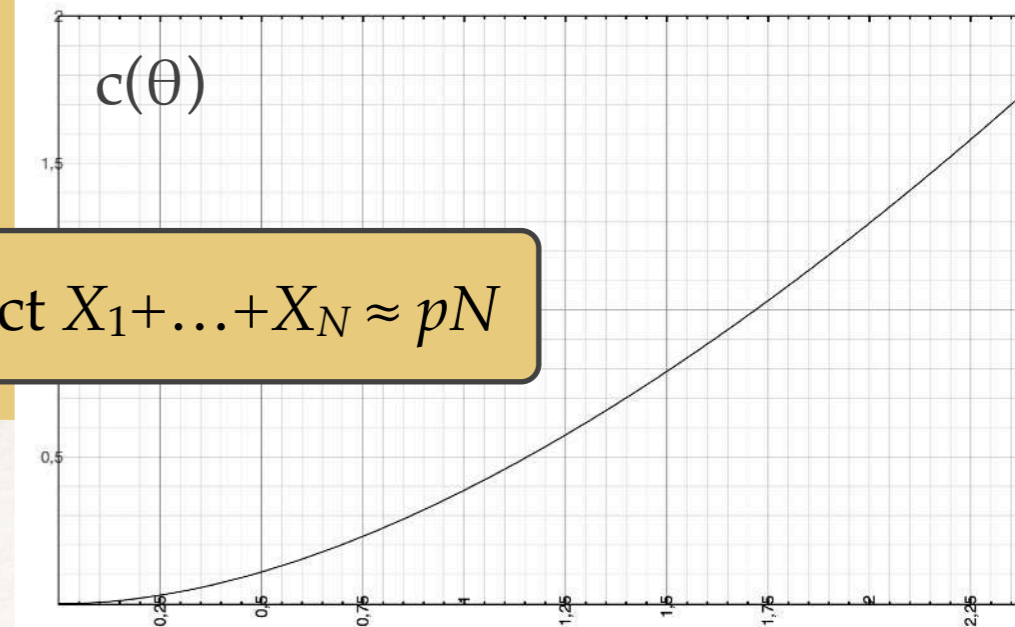
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We expect $X_1 + \dots + X_N \approx pN$

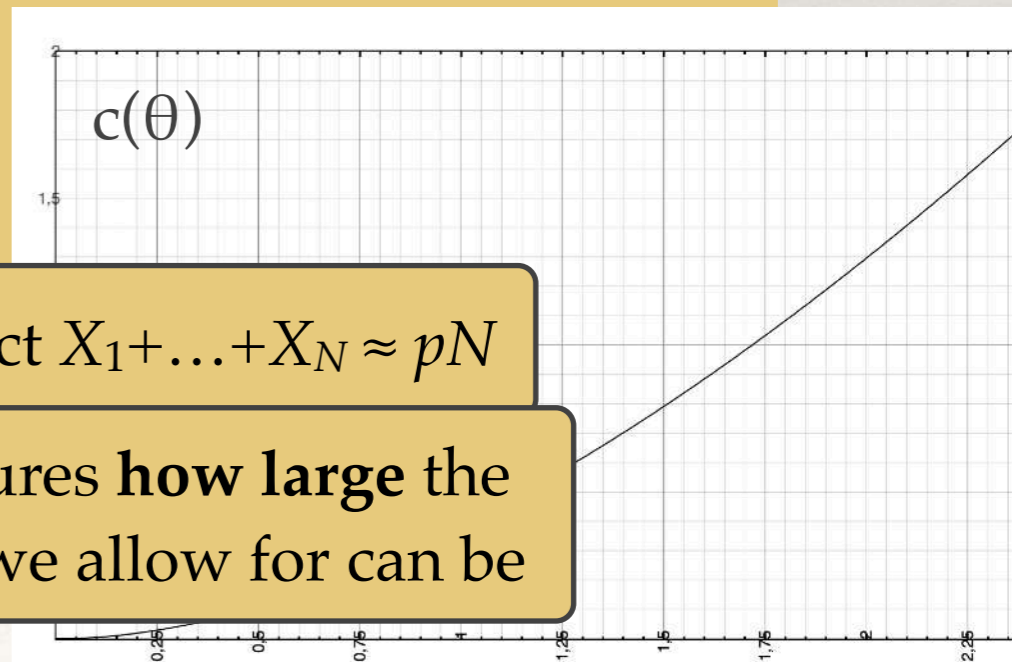


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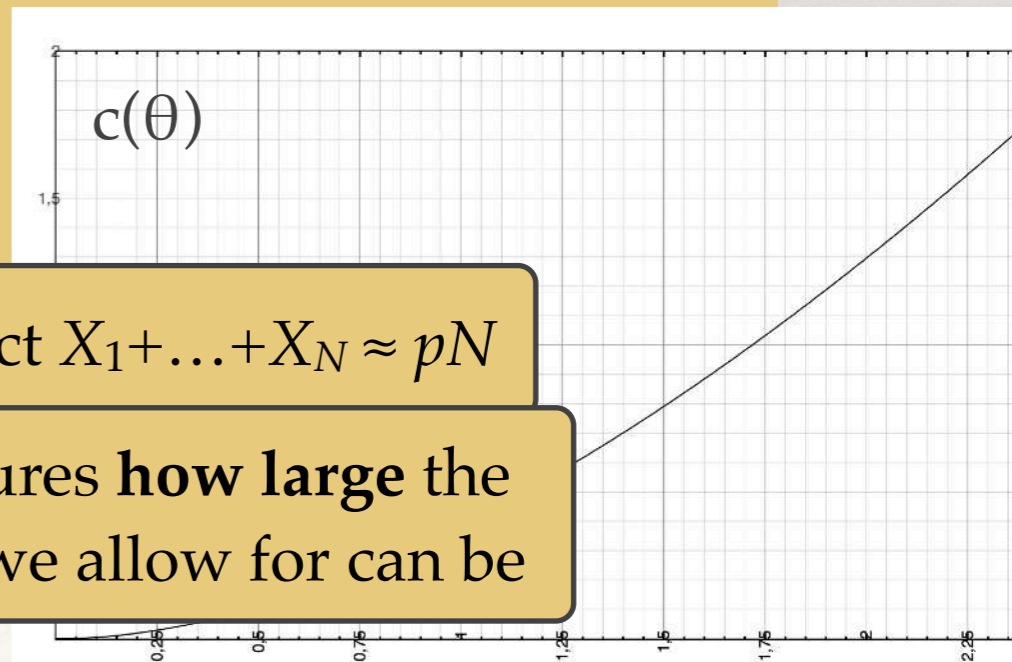
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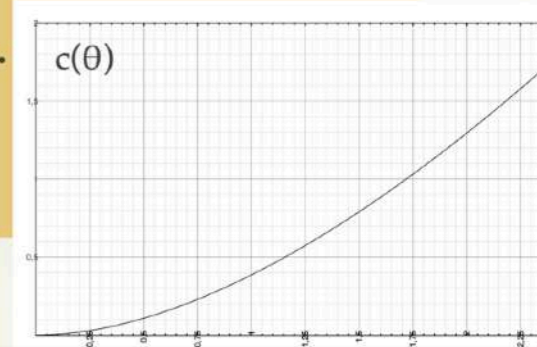
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For all practical purposes, $c(\theta) \approx \theta^2/3$

Proof of Chernoff's bound (1/4)

- ❖ Let $t, a > 0$ to be fixed later
- ❖ Define the rand. var
 $X = \exp(t(X_1 + \dots + X_N))$

Theorem. Let X_1, \dots, X_N be independent rand. vars with values in $\{0, 1\}$ and with the same law: $\Pr(X_i=1)=p$. Then $\Pr(X_1 + \dots + X_N \geq (1+\theta)pN) \leq \exp(-c(\theta)pN)$

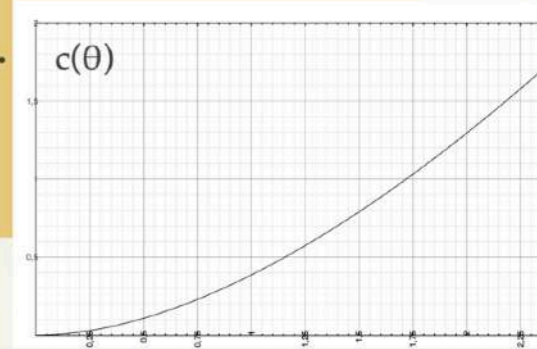


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- ❖ Define the rand. var
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- ❖ Note that $E(X) \leq \exp(tN) < \infty$, so we can use **Markov's inequality**:

$$\Pr(X \geq a \cdot E(X)) \leq 1/a$$

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Theorem (Markov's inequality).

Let X be a **non-negative real-valued** random variable with **finite** expectation $E(X)$. For every $a \geq 0$:
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Proof of Chernoff's bound (2/4)

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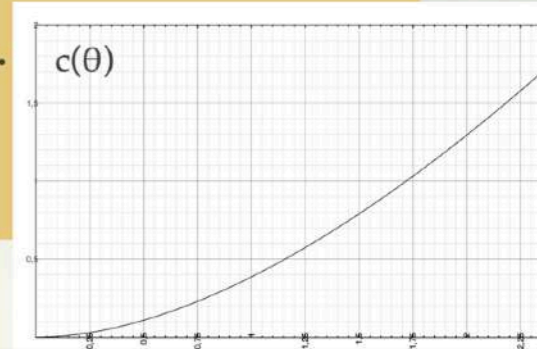
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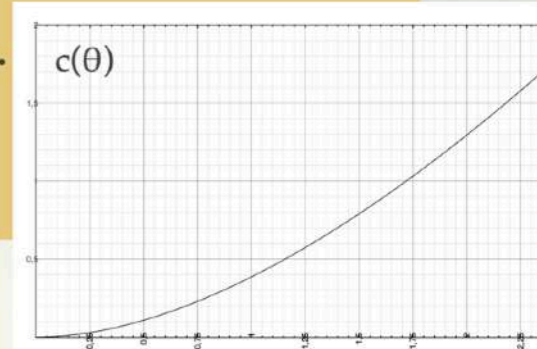
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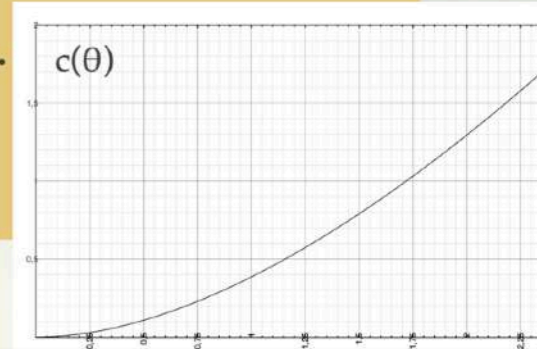
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This is just $\Pr(X_1 + \dots + X_N \geq (1+\theta)pN)$

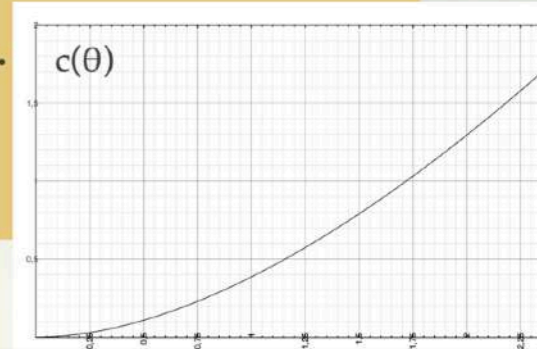
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Proof of Chernoff's bound (3/4)

- ❖ Let $t > 0$, to be fixed later
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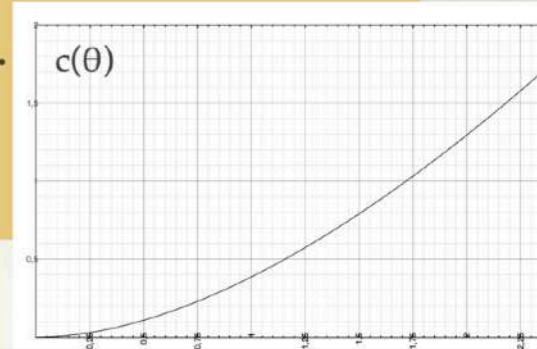
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- ❖
$$\begin{aligned} E(X) &= E(\prod_{i=1}^N \exp(tX_i)) \\ &= \prod_{i=1}^N E(\exp(tX_i)) && \text{(independence)} \\ &= \prod_{i=1}^N (p \exp(t) + 1 - p) && \text{(def. of the law of } X_i) \\ &= (p \exp(t) + 1 - p)^N \\ &= (1 + p(\exp(t) - 1))^N \leq \exp((\exp(t) - 1)pN) \end{aligned}$$

Theorem. Let X_1, \dots, X_N be independent rand. vars with values in $\{0, 1\}$ and with the same law: $\Pr(X_i=1)=p$. Then $\Pr(X_1 + \dots + X_N \geq (1 + \theta)pN) \leq \exp(-c(\theta)pN)$



take logs:

$$N \log(1 + p(\exp(t) - 1)) \leq Np(\exp(t) - 1)$$

Proof of Chernoff's bound (4/4)

❖ Let $t > 0$, to be fixed later

❖ $X = \exp(t(X_1 + \dots + X_N))$

❖ $\Pr(X_1 + \dots + X_N \geq (1 + \theta)pN)$

$$\leq \exp((\exp(t) - 1)pN) \exp(-t(1 + \theta)pN)$$

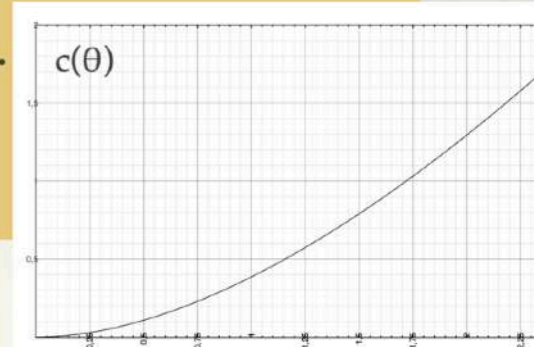
(from last slide)

Theorem. Let X_1, \dots, X_N be independent rand. vars with values in $\{0, 1\}$ and

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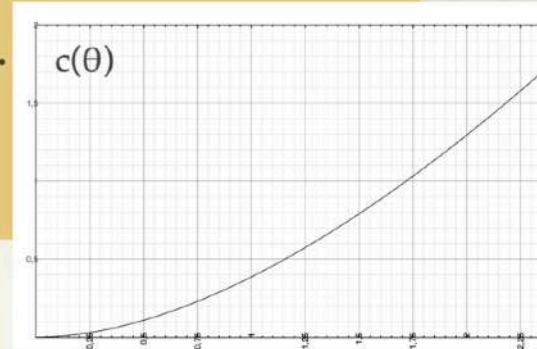
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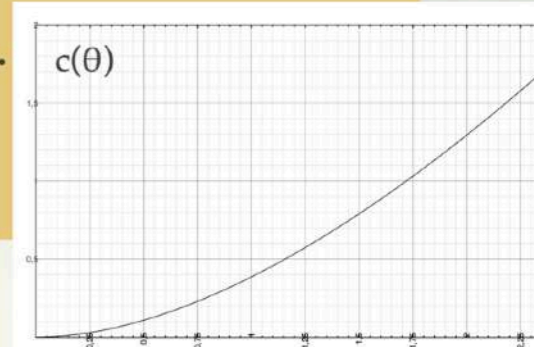
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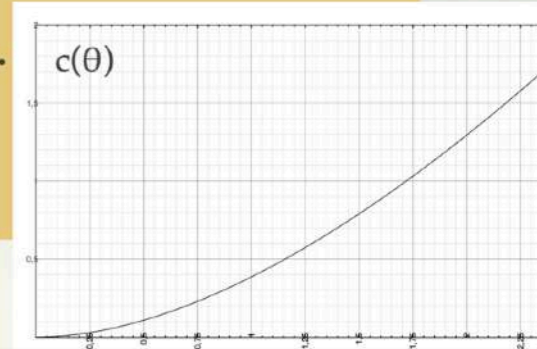
❖ Let $t = \log(1 + \theta)$, so $(\exp(t) - 1)pN = \theta pN$, hence

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Done! \square

Call this $-c(\theta)$

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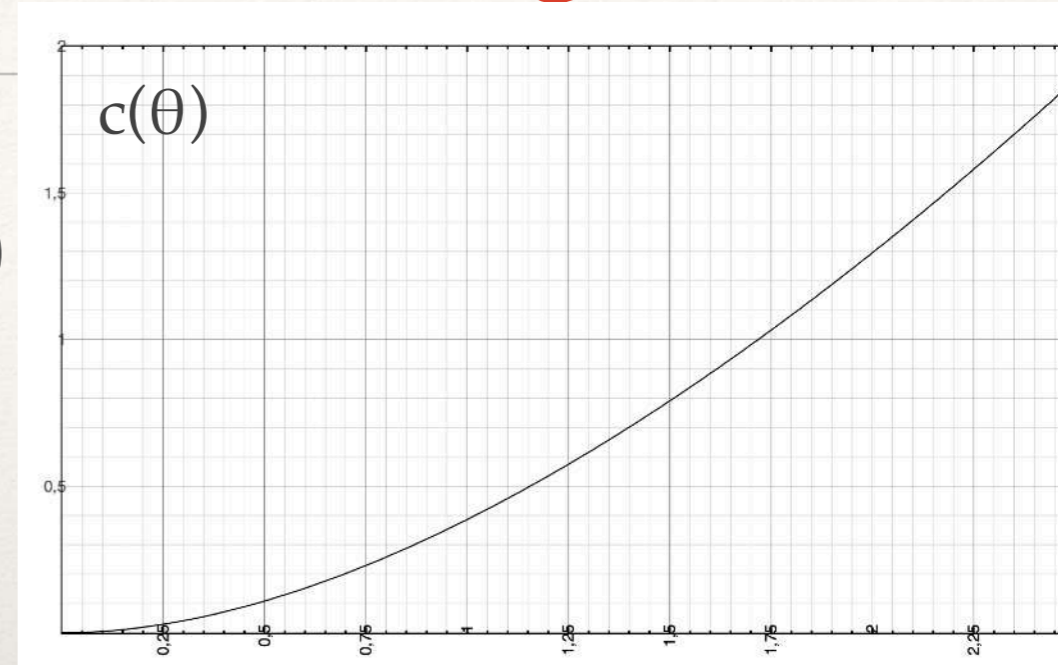


A few properties of $c(\theta) = -\theta + (1+\theta)\log(1+\theta)$

- ❖ **Prop 1.** $c(\theta)$ is monotonic (for $\theta \geq 0$)
- ❖ *Proof.* $c'(\theta) = \log(1+\theta) \geq 0$

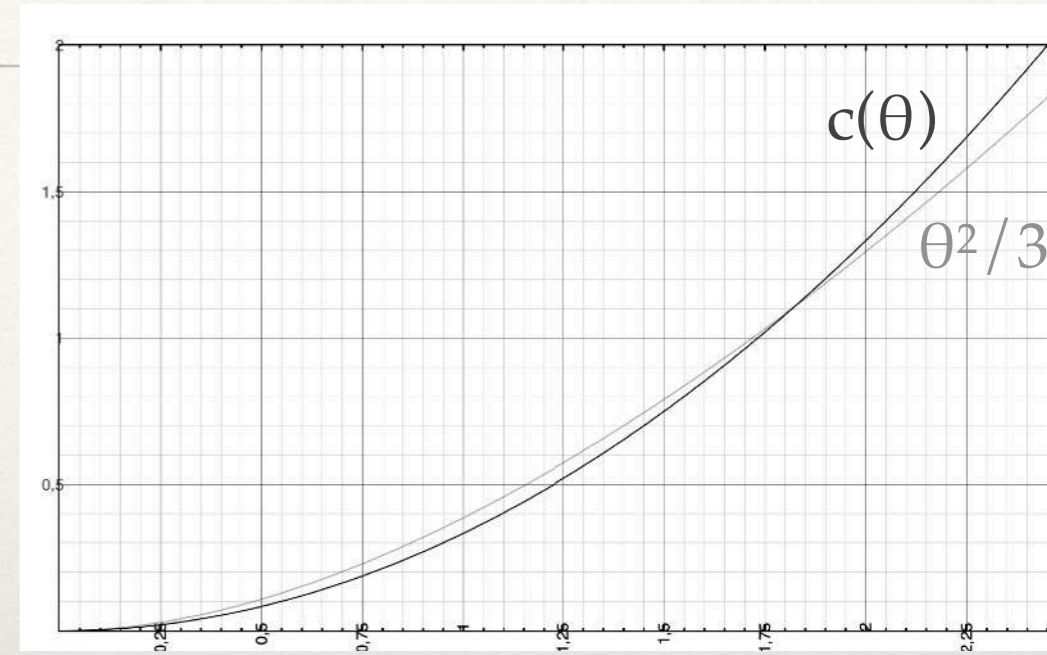
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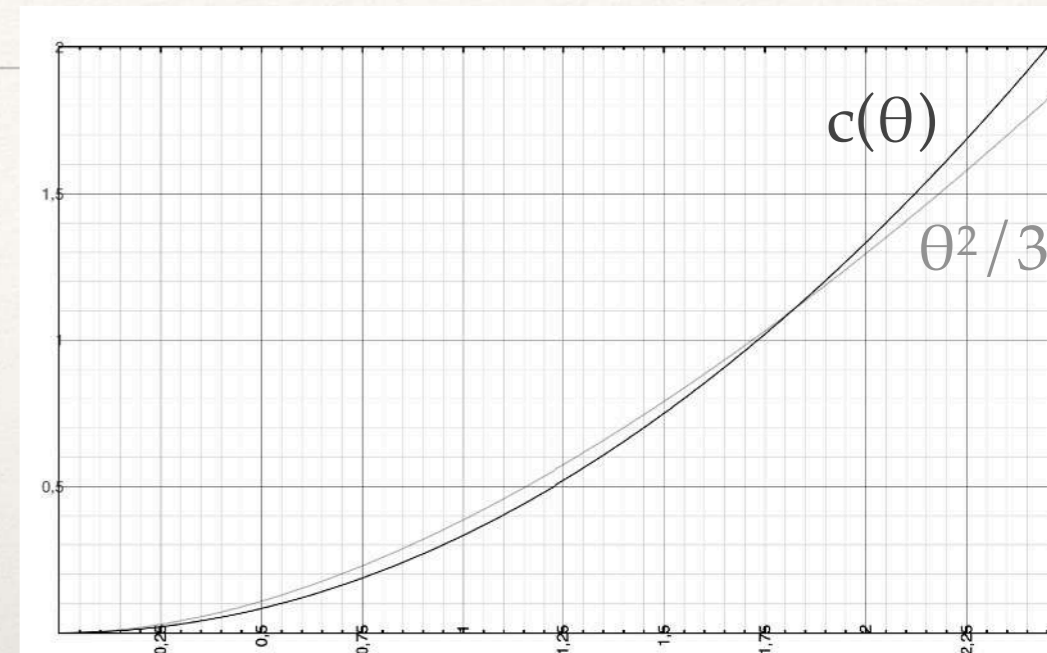
❖ **Prop 2.** For $0 \leq \theta \leq 1$, $c(\theta) \geq \theta^2/3$

❖ *Proof.* $c(0) = 0$

$$c'(0) = 0 \quad (\text{recall } c'(\theta) = \log(1+\theta))$$

$$c''(0) = 1 \quad (c''(\theta) = 1/(1+\theta))$$

$$c'''(0) = -1 \quad (c'''(\theta) = -1/(1+\theta)^2)$$



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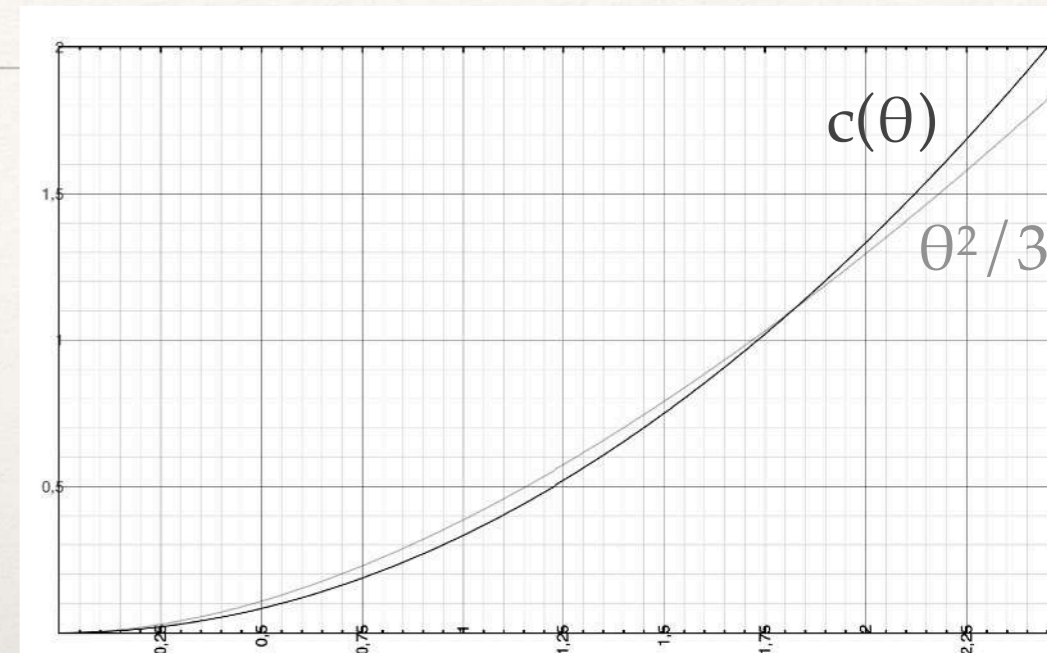
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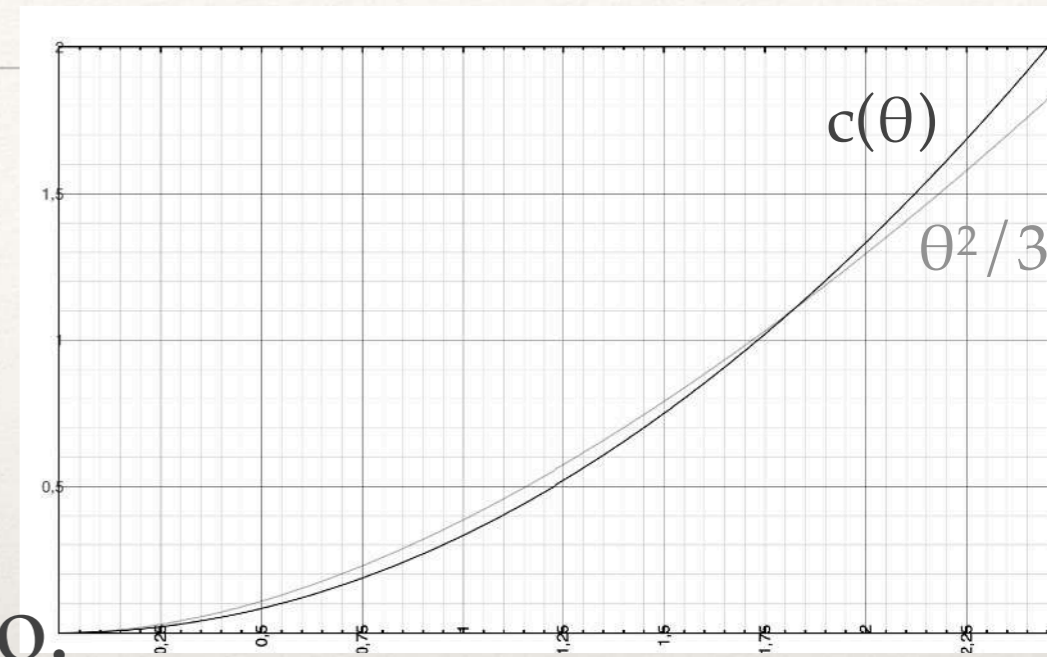
$$c'''(0) = -1 \quad (c'''(\theta) = -1/(1+\theta)^2)$$

❖ So $c(\theta) = \theta^2/2 - \theta^3/6 + c^{(4)}(\theta_0)/24$ for some $0 \leq \theta_0 \leq \theta$ (Taylor)

$$\geq \theta^2/2 - \theta^3/6 \quad (\text{since } c^{(4)}(\theta) = 2/(1+\theta)^3 \geq 0)$$
$$\geq \theta^2/3 \quad (\text{since } \theta \leq 1) \quad \square$$


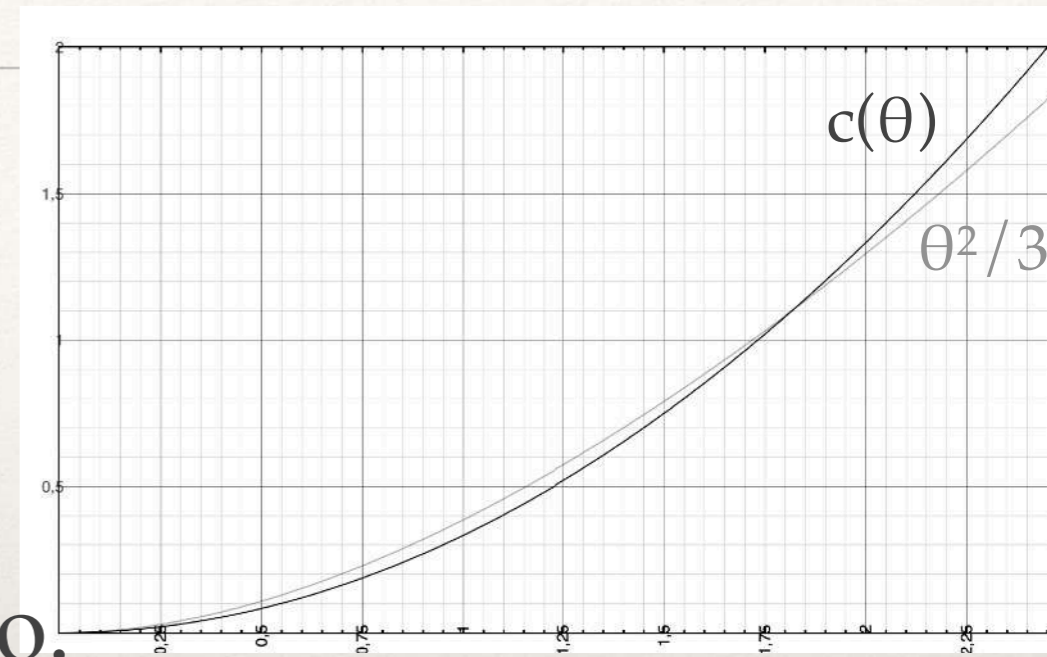
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- ❖ **Proof.** $c(\theta)/(1+\theta) = -\theta/(1+\theta) + \log(1+\theta)$
Derivative: $-1/(1+\theta)^2 + 1/(1+\theta) = \theta/(1+\theta)^2 \geq 0 \quad \square$



Application to voting (1/4)

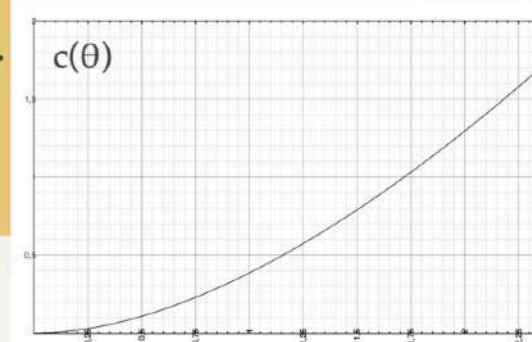
- ❖ Assume that $\Pr_r(\mathcal{M}(x,r) \text{ errs}) \leq 1/3$,
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all assumptions satisfied
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(Chernoff)

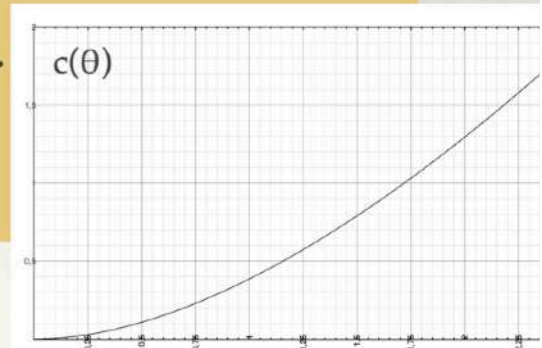


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Application to voting (2/4)

- ❖ Assume that $\Pr_r(\mathcal{M}(x,r) \text{ errs}) \leq 1/3$,
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(from last slide)
- ❖ I.e., $P \leq \exp(-c(\theta)/(1+\theta) \cdot 1/2 N)$

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- ❖ I.e., $P \leq \exp(-c(\theta)/(1+\theta) \cdot 1/2 N)$
- ❖ $\leq \exp(-c(1/2)/(3/2) \cdot 1/2 N)$
(since $p \leq 1/3$, so $\theta \geq 1/2$; plus Prop 3)

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- ❖ $\leq \exp(-c(1/2)/(3/2) \cdot 1/2 N)$
(since $p \leq 1/3$, so $\theta \geq 1/2$; plus Prop 3)
- ❖ $\leq \exp(-(1/2)^2/3 / (3/2) \cdot 1/2 N)$ (Prop 2)
 $= \exp(-N/36)$

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Application to voting (3/4)

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- ❖ Answer: at most $\exp(-N/36)$

Error reduction for BPP

- ❖ First, a useful trick.
Let us say that $\mathcal{M}(x,r)$ **errs**
iff $(x \in L$ and $\mathcal{M}(x,r)$ rejects)
or $(x \notin L$ and $\mathcal{M}(x,r)$ accepts)
- ❖ (That used to be implicit.)
- ❖ Then:

A language L is in **BPP** if and only if there is a **polynomial-time** TM \mathcal{M} such that for every input x (of size n):
if $x \in L$ then $\Pr_r [\mathcal{M}(x,r) \text{ accepts}] \geq 2/3$
if $x \notin L$ then $\Pr_r [\mathcal{M}(x,r) \text{ accepts}] \leq 1/3$.

A language L is in **BPP(ϵ)** and only if there is a **polynomial-time** TM \mathcal{M} such that for every input x (of size n):
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error = ϵ

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Error reduction for BPP

- ❖ Let L be in **BPP**, as here \rightarrow
- ❖ Build new rand. TM \mathcal{M}' by:
- ❖ $yeas := 0$
for $i=1$ to N :
 draw r at random
 if $\mathcal{M}(x,r)$ accepts:
 $yeas++$
accept if $yeas \geq N/2$, else reject

A language L is in **BPP** if and only if there is a **polynomial-time** TM \mathcal{M} such that for every input x (of size n):

$$\Pr_r (\mathcal{M}(x,r) \text{ errs}) \leq 1/3.$$

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 - if $\mathcal{M}(x,r)$ accepts:
 - $yeas++$
 - accept if $yeas \geq N/2$, else reject
- ❖ \mathcal{M}' errs on input x iff at least half of the calls to $\mathcal{M}(x,r)$ err
- ❖ That happens with probability $\leq \exp(-N/36)$
- ❖ $\dots \leq \varepsilon$ provided that we pick $N \geq -36 \log \varepsilon$

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Note: if \mathcal{M} runs in polytime $p(n)$,
then \mathcal{M}' runs in **polytime** $= -36 \log \varepsilon p(n) + \text{cst.}$

Error reduction for BPP

- ❖ Hence $\mathbf{BPP}(= \mathbf{BPP}(1/3)) \subseteq \mathbf{BPP}(\varepsilon)$
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- ❖ **Theorem.** For every ε , $0 < \varepsilon < 1/2$, $\mathbf{BPP} = \mathbf{BPP}(\varepsilon)$.
- ❖ ... but can we do better?

Application to voting (4/4)

- ❖ Assume that $\Pr_r(\mathcal{M}(x,r) \text{ errs}) \leq 1/3$,
how large should N be so that
the probability P that more than $1/2$
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- ❖ *Proof.* $\exp(-N/36) \leq 1/2^{q(n)}$ iff
 $-N/36 \leq -q(n) \log 2$

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- ❖ Answer: at least $36 q(n) \log 2$

- ❖ *Proof.* $\exp(-N/36) \leq 1/2^{q(n)}$ iff
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Application to voting (3/4)

- ❖ Assume that $\Pr_r(\mathcal{M}(x,r) \text{ errs}) \leq 1/3$,
what is the probability P that more than $1/2$ of N
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The only magical formula
you'll need to remember
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Application to voting (4/4)

- ❖ Assume that $\Pr_r(\mathcal{M}(x,r) \text{ errs}) \leq 1/3$,
how large should N be so that
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Note: if $q(n)$ is polynomial,
this is polynomial, too

Error reduction for BPP revisited

- ❖ Let L be in **BPP**
- ❖ Build new rand. TM \mathcal{M}' by:
 - ❖ $\text{yeas} := 0$
 - ❖ for $i=1$ to $N := 36 q(n) \log 2$:
 - draw r at random
 - if $\mathcal{M}(x,r)$ accepts:
 - $\text{yeas}++$
 - ❖ **accept** if $\text{yeas} \geq N/2$, else **reject**
- ❖ \mathcal{M}' errs on input x iff at least half of the calls to $\mathcal{M}(x,r)$ err
- ❖ That happens with probability $\leq 1 / 2^{q(n)}$

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Note: if \mathcal{M} runs in polytime $p(n)$, and $q(n)$ is polynomial then \mathcal{M}' runs in **polytime** = $O(q(n) p(n) \log n)$ [$\log n$ for operations on the counter i]

Error reduction for BPP revisited

A language L is in **BPP**(ϵ) and only if there is a **polynomial-time** TM \mathcal{M} such that for every input x (of size n):

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error = ϵ

- ❖ **Theorem.** BPP is equal to:
 - BPP(ϵ) for every $\epsilon, 0 < \epsilon < 1/2$
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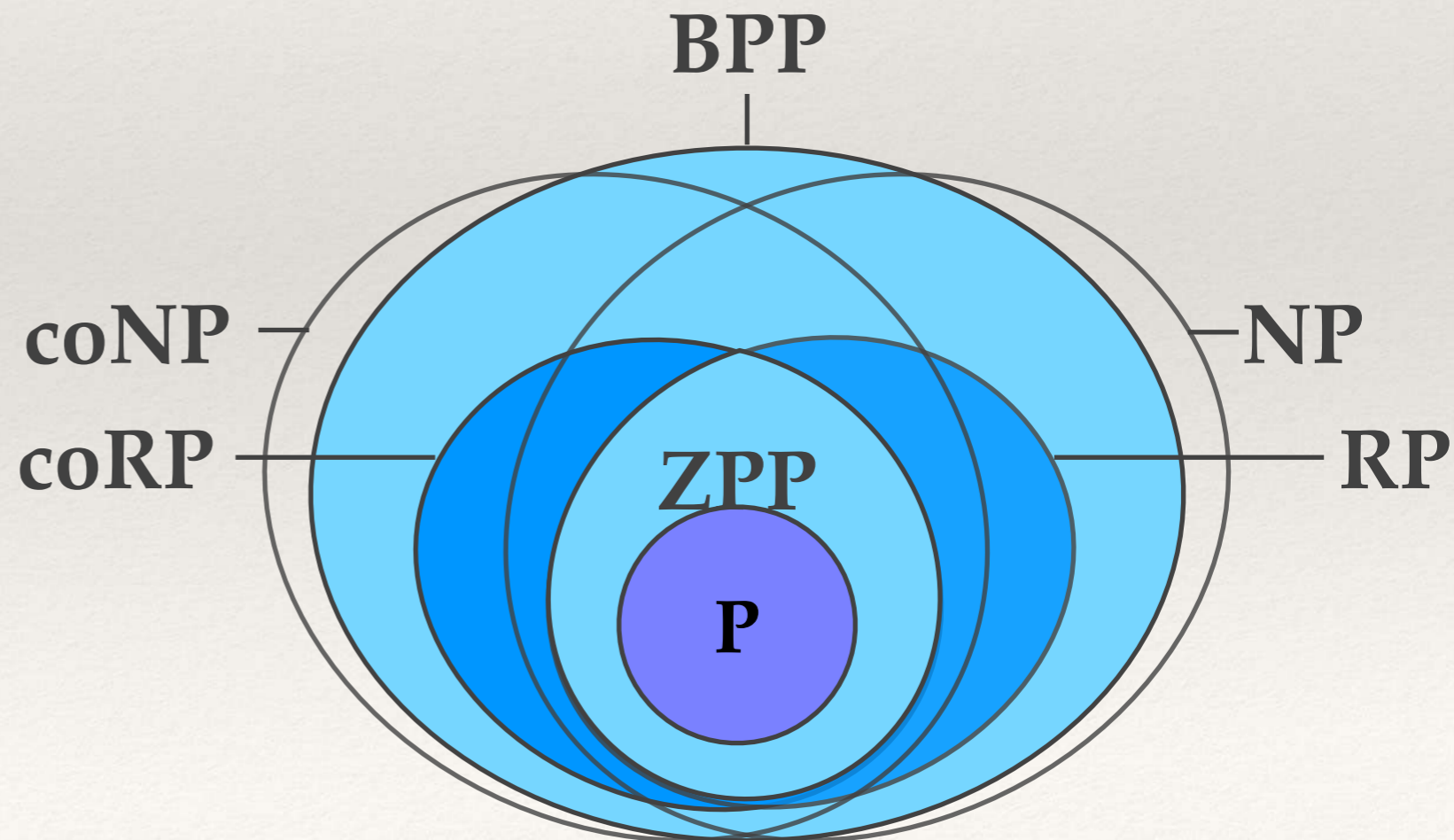
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The new landscape

BPP vs. other complexity classes

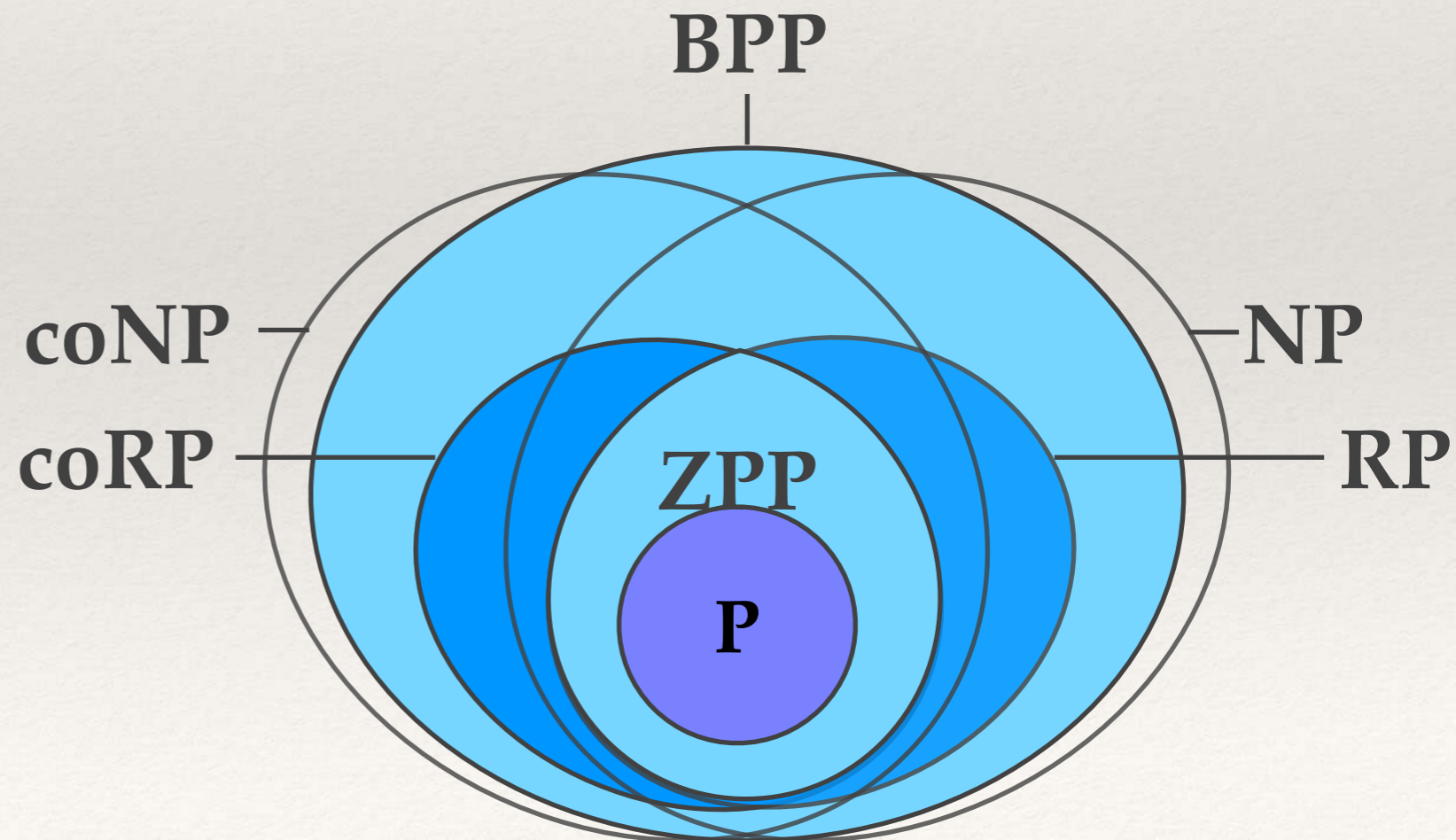
- ❖ Both **RP** and **coRP** are included in **BPP**
(if you make a mistake with prob. 0, then this prob. is $\leq 1/3!$)
- ❖ **BPP** is closed under complements: **BPP=coBPP**
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BPP vs. other complexity classes

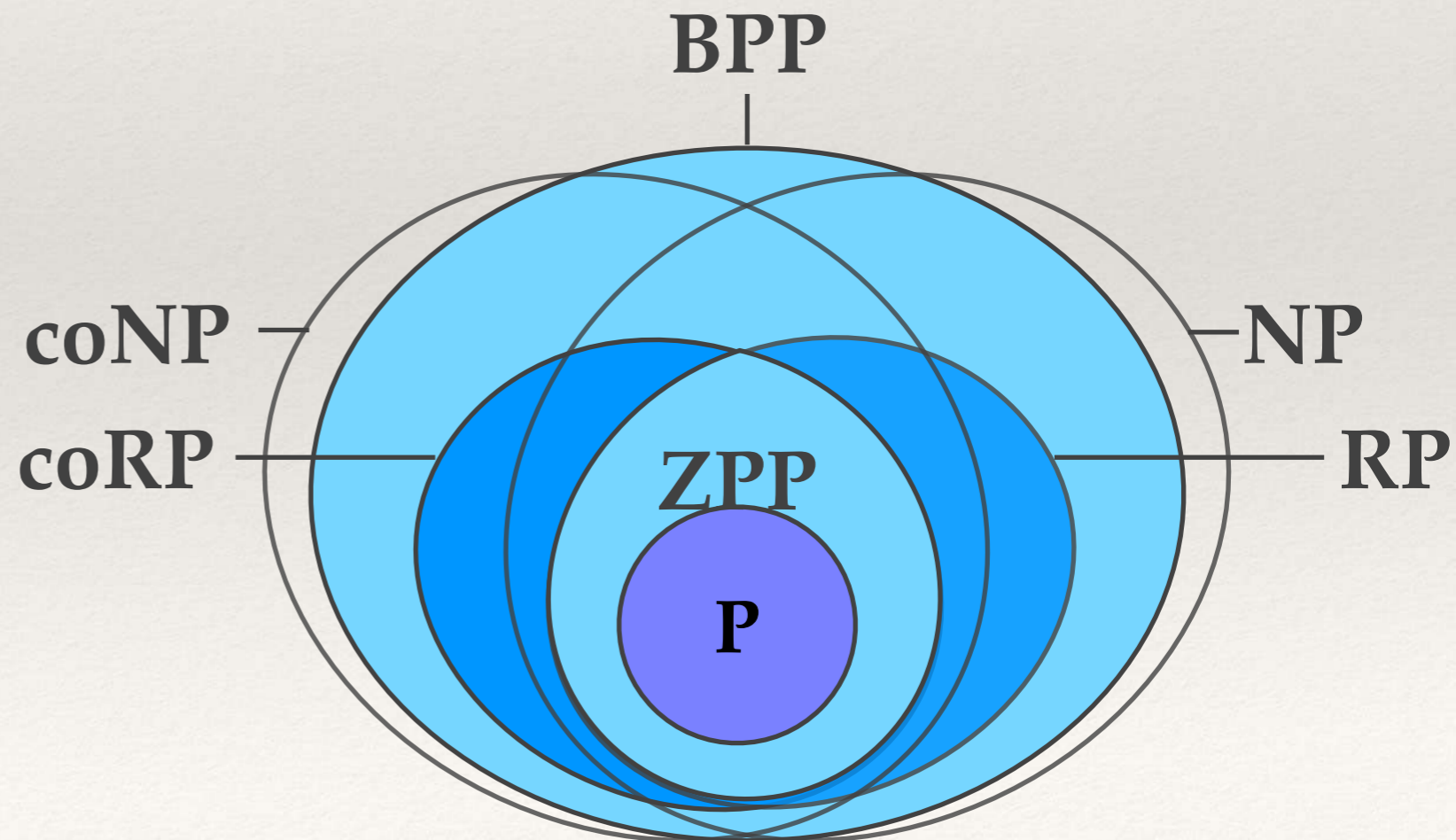
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- ❖ ... but what is the relation between **BPP**, **NP**, **coNP**, etc.?



BPP cannot be too large

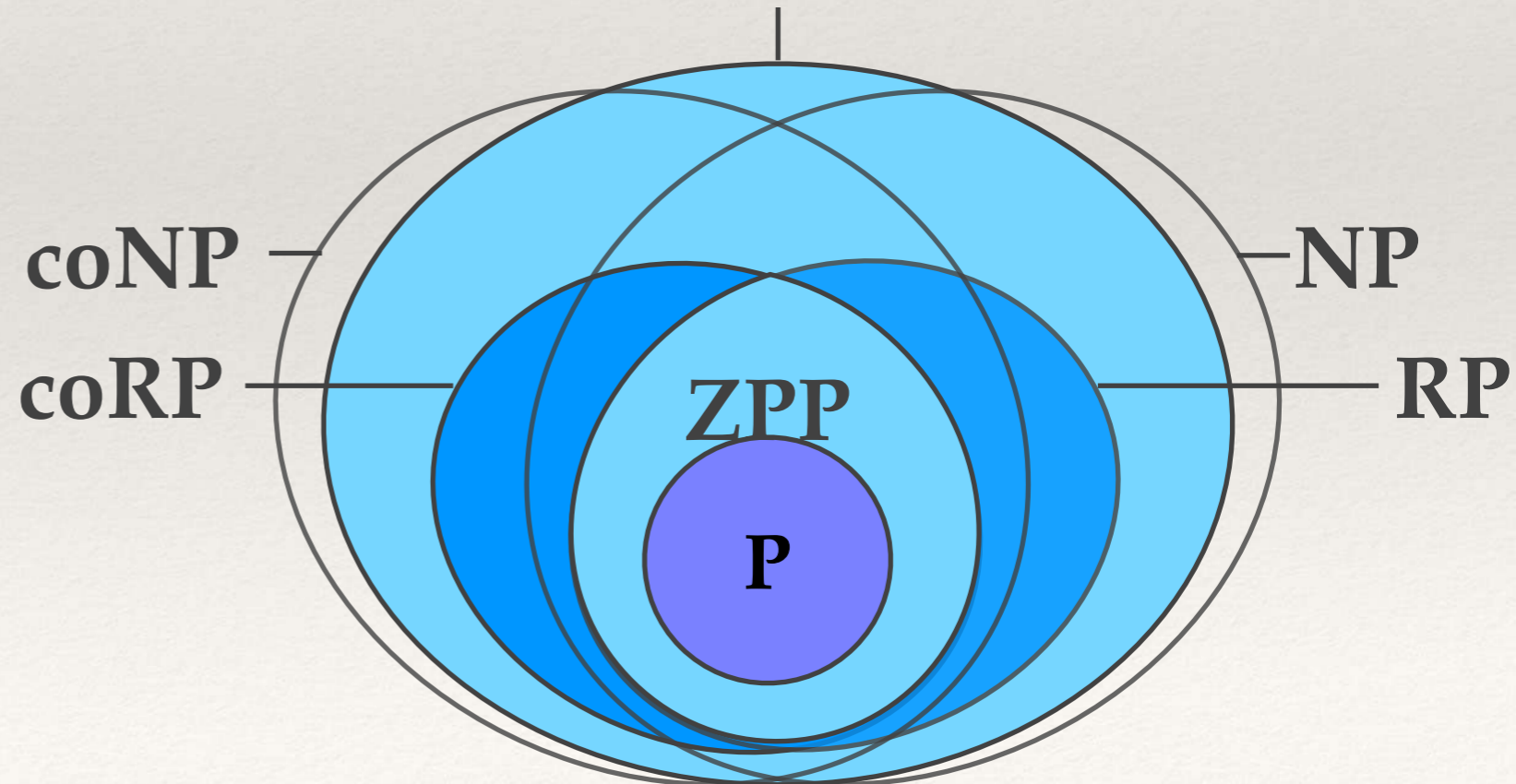
- ❖ It is unknown whether $\mathbf{BPP} \subseteq / \supseteq \mathbf{NP}$ (eqv., \mathbf{coNP})
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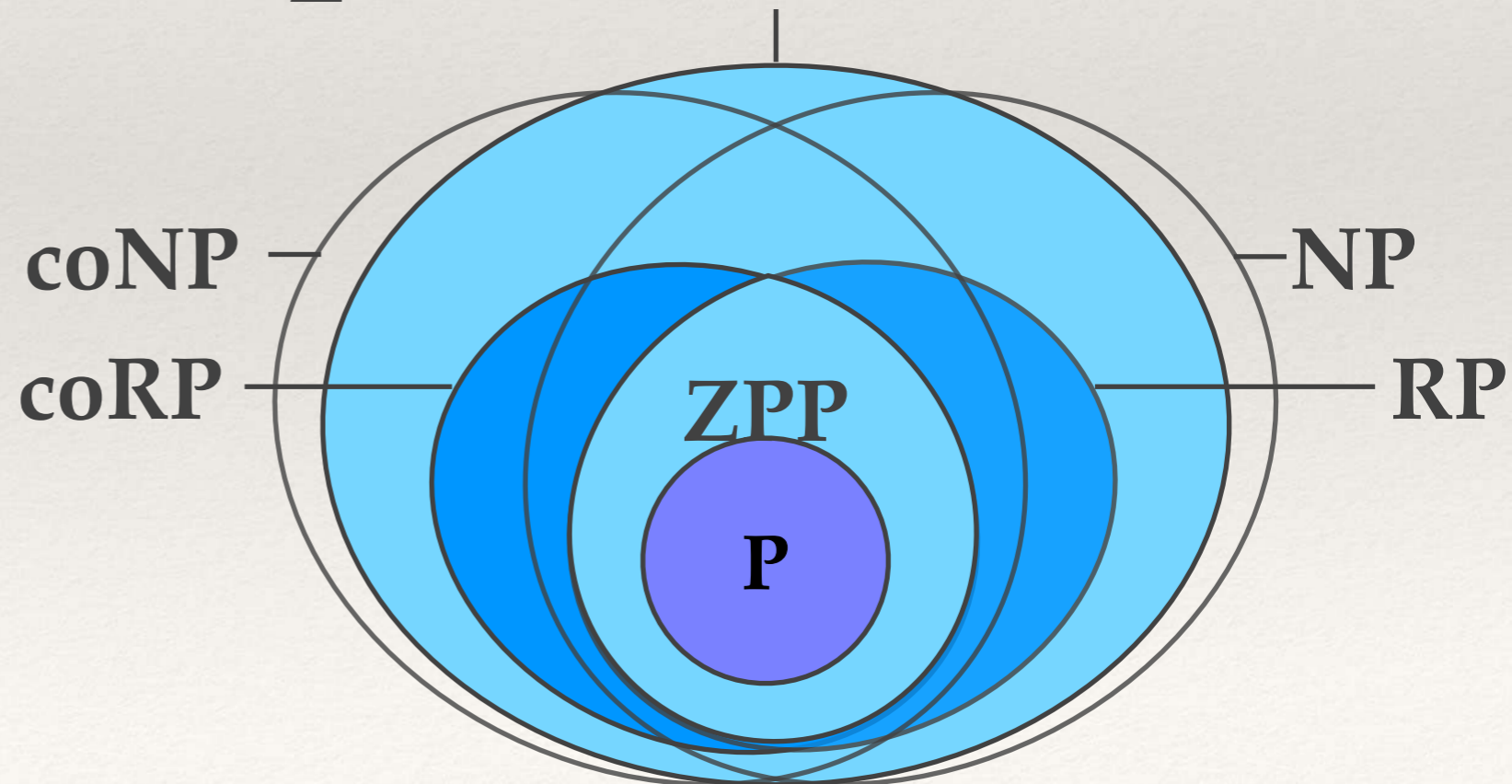


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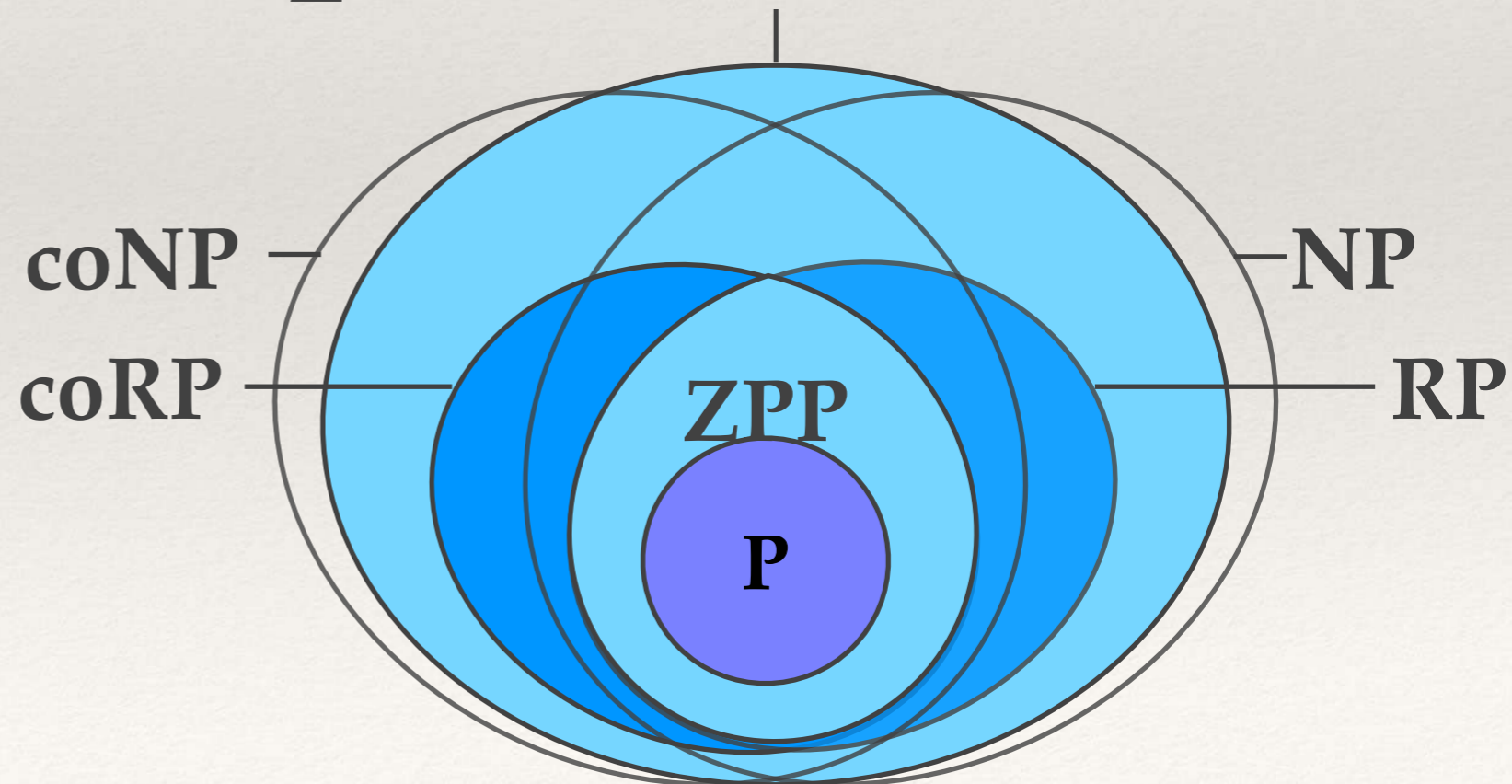
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The Sipser-Gács-Lautemann Theorem



http://lpcs.math.msu.su/~ver/photo_album/Colleagues/lautemann+allender+wagn

<https://gravatar.com/avatar/dc36e666740ff9480eb738e556c887a4?s=200>

https://upload.wikimedia.org/wikipedia/commons/thumb/3/34/MIT-Science_Sipser_Michael.jpg/440px-MIT-Science_Sipser_Michael.jpg

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❖ **Theorem (Sipser-Gács-Lautemann, Prop. 1.24.)**

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❖ *Proof sketch.*

It is enough to prove $\mathbf{BPP} \subseteq \Sigma^{\text{P}_2}$.

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In order to do so, we will to prove
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To prove that $\exists t, P(t)$,
just show that $\Pr_t(P(t)) \neq 0$, or
equivalently that $\Pr_t(\neg P(t)) < 1$

Lautemann's trick

- ❖ Let $L \in \mathbf{BPP}$, decided with error $\varepsilon = 1/2^n$ (not $1/3$) in polytime $p(n)$

A language L is in $\mathbf{BPP}(\varepsilon)$ and only if there is a **polynomial-time** TM \mathcal{M} such that for every input x (of size n):

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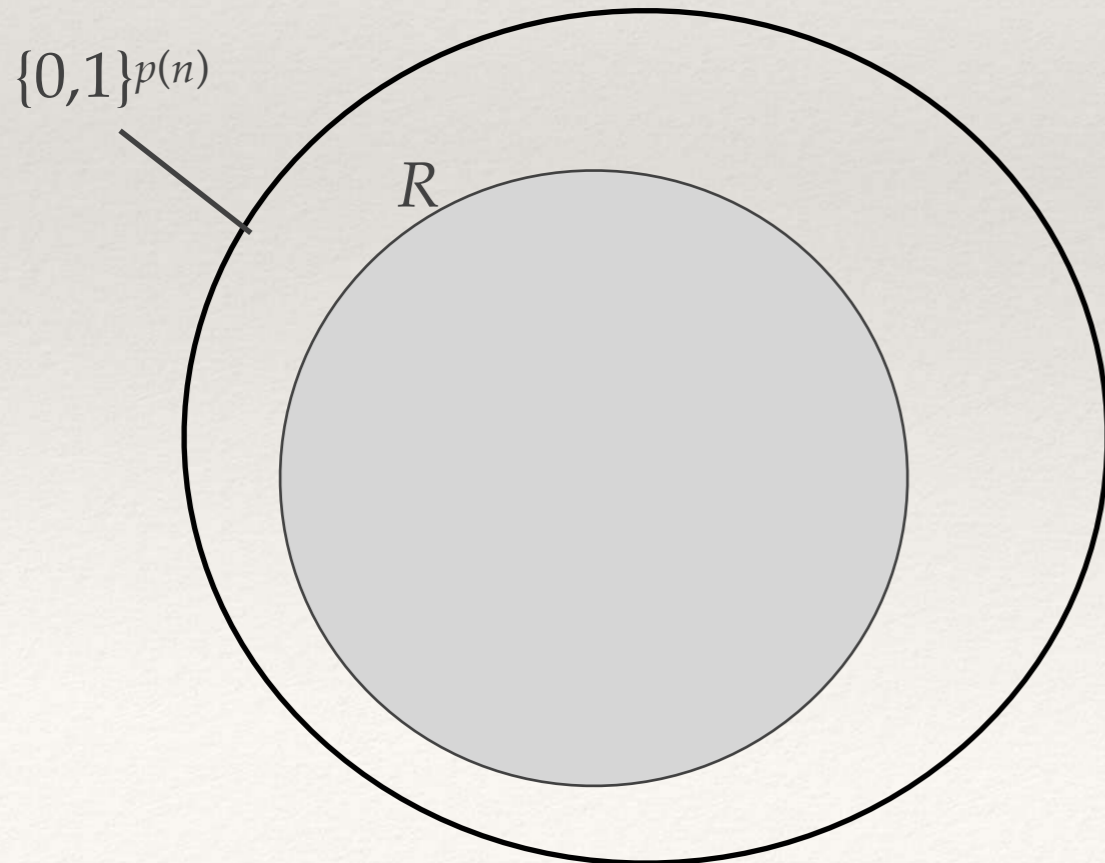
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either **huge**, if $x \in L$
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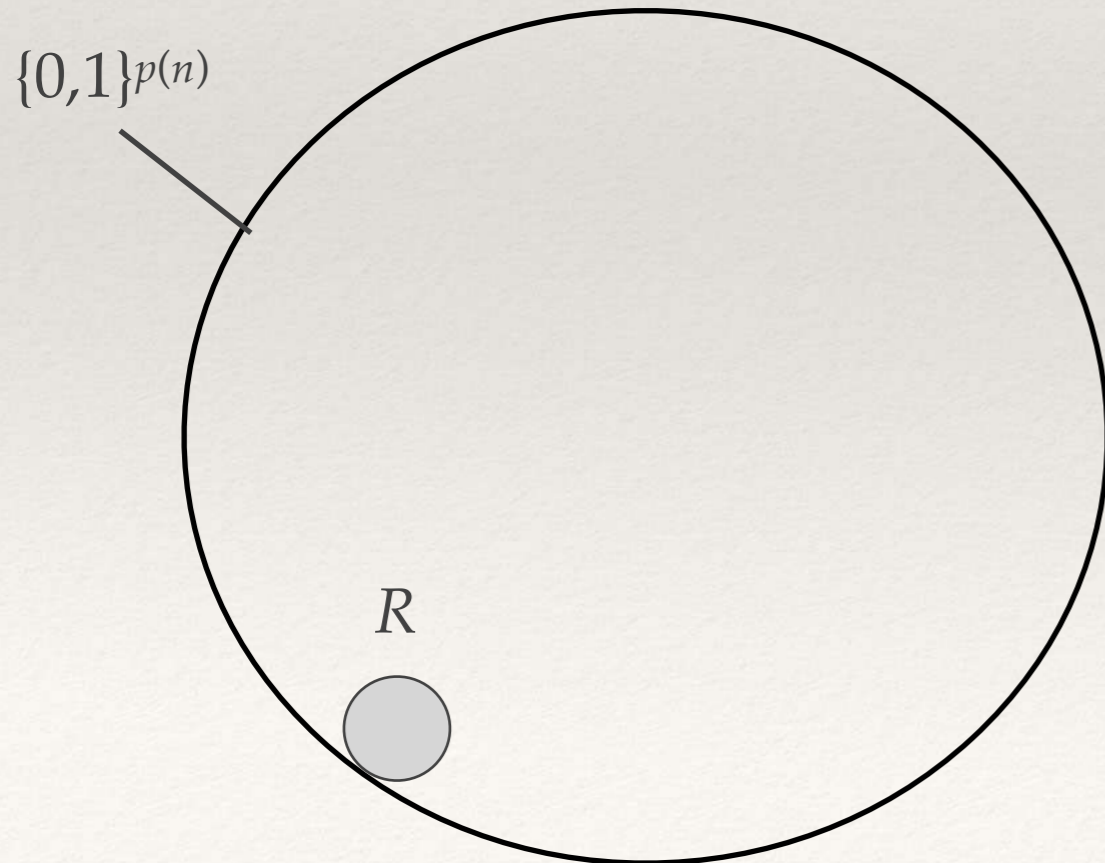
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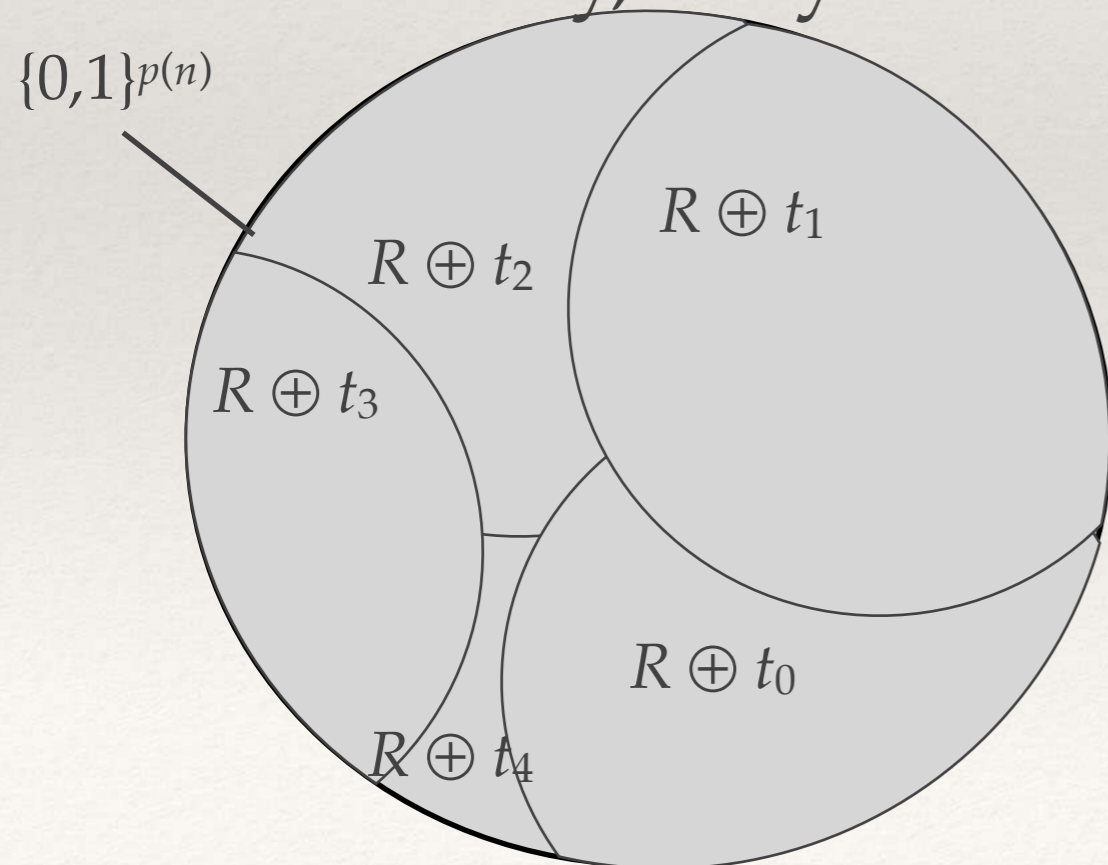


error $\varepsilon = 1/2^n$

or **tiny**, if $x \notin L$
(covers a proportion $\leq 1/2^n$
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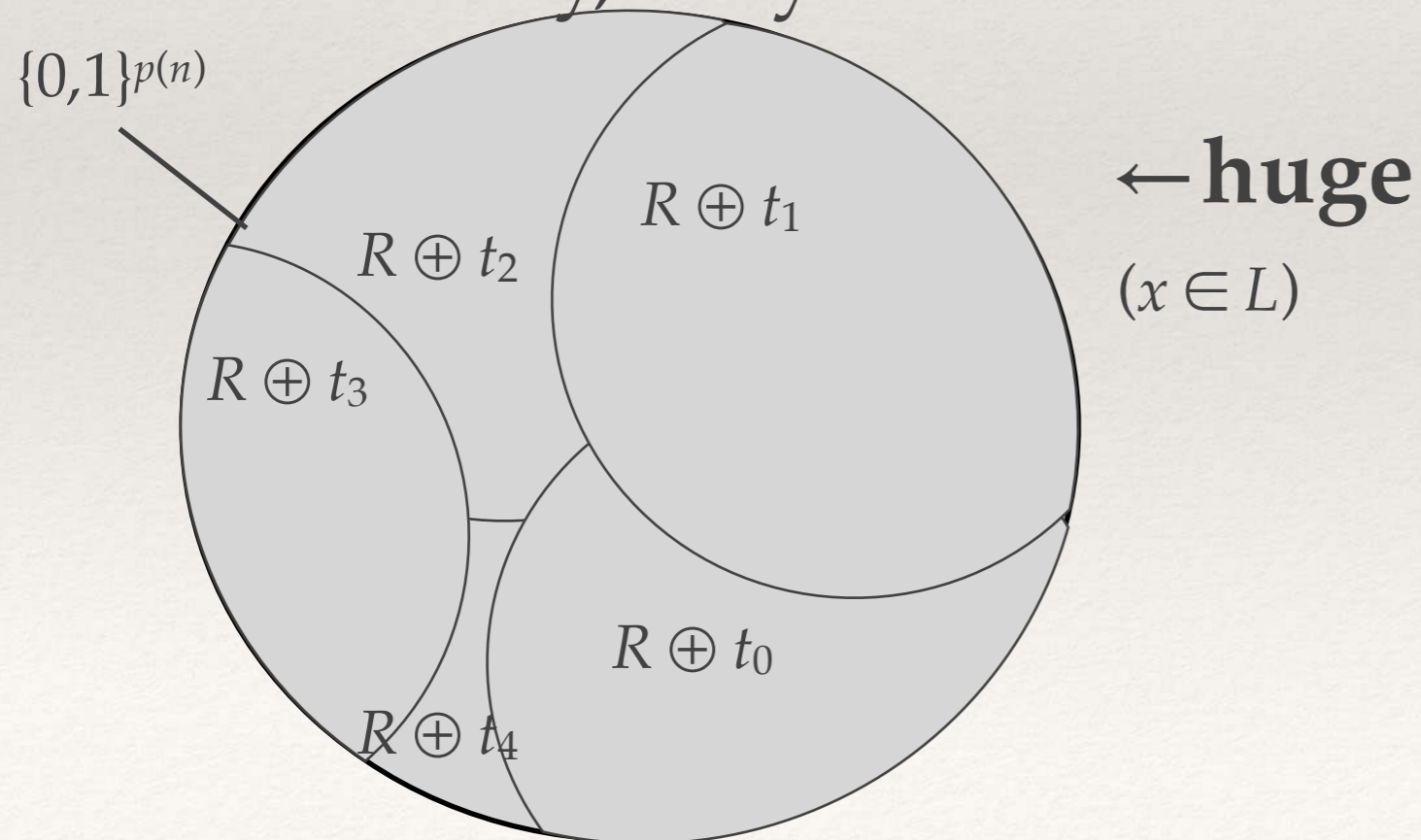
Lautemann's trick

- ❖ $R = \{r \in \{0,1\}^{p(n)} \mid \mathcal{M}(x,r) \text{ accepts}\}$ is **huge** or **tiny**
- ❖ We claim there are **translations** $R \oplus t_i$ of R such that:
 - if R huge, then the translations cover the whole space
 - if R tiny, they do not.



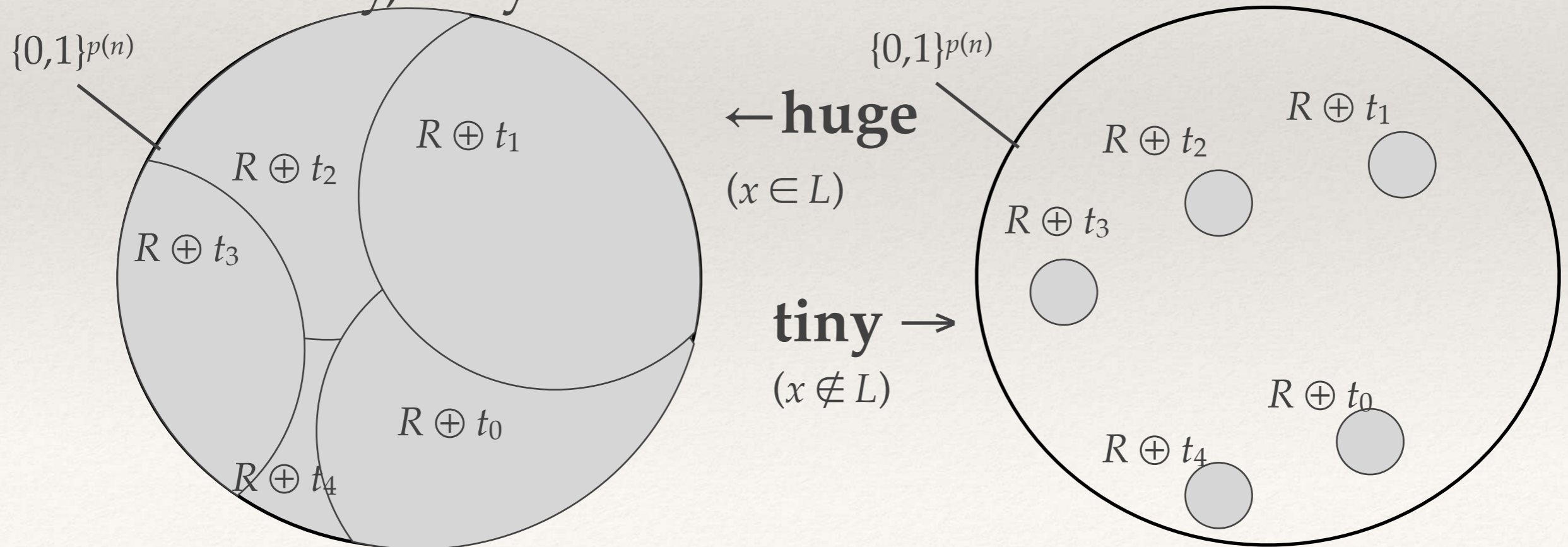
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Translations?

- ❖ The computer science view:

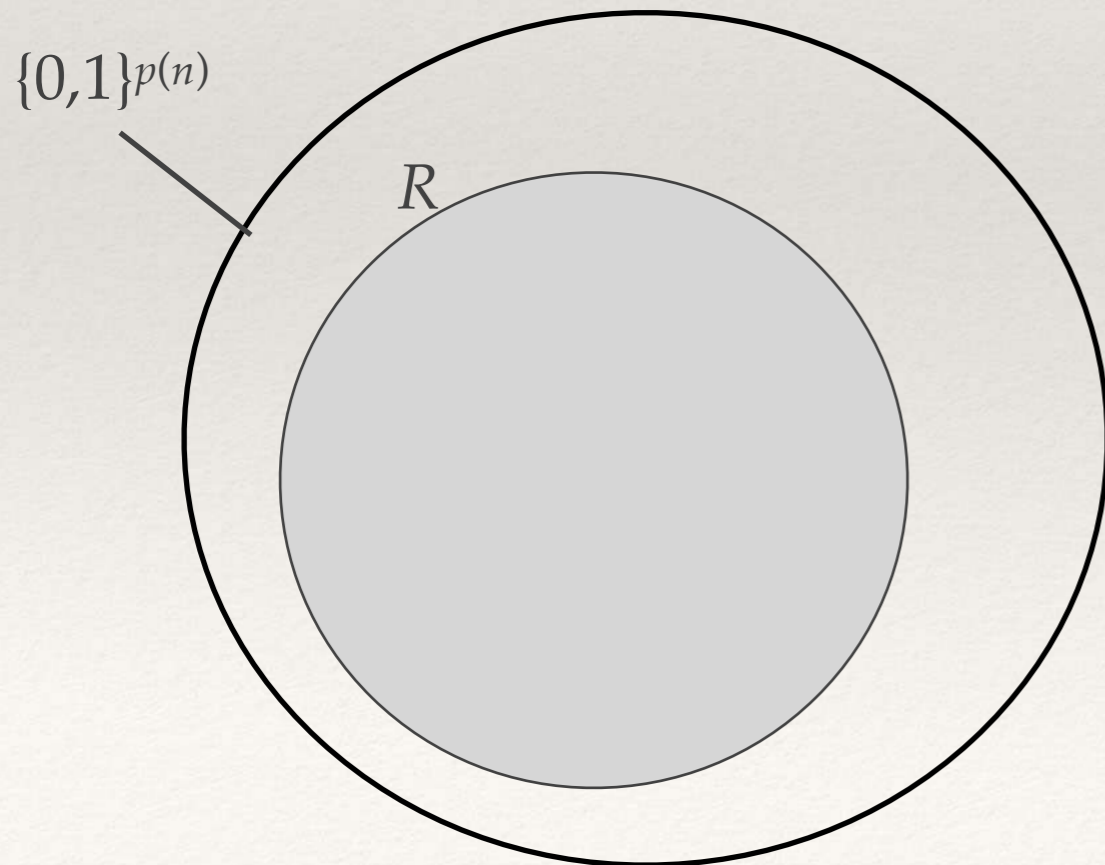
\oplus is **bitwise exclusive-or**

r	0	0	0	0	1	0	1	0	0	0	0	1	0	1	0	0	1	0	0	0	0	1	1
t	1	0	1	0	1	1	0	1	0	1	0	1	1	0	1	0	0	1	0	1	0	1	1
$r \oplus t$	1	0	1	0	0	1	1	1	0	1	0	0	1	1	1	0	1	1	0	1	0	0	0

- ❖ $R \oplus t = \{r \oplus t \mid r \in R\}$

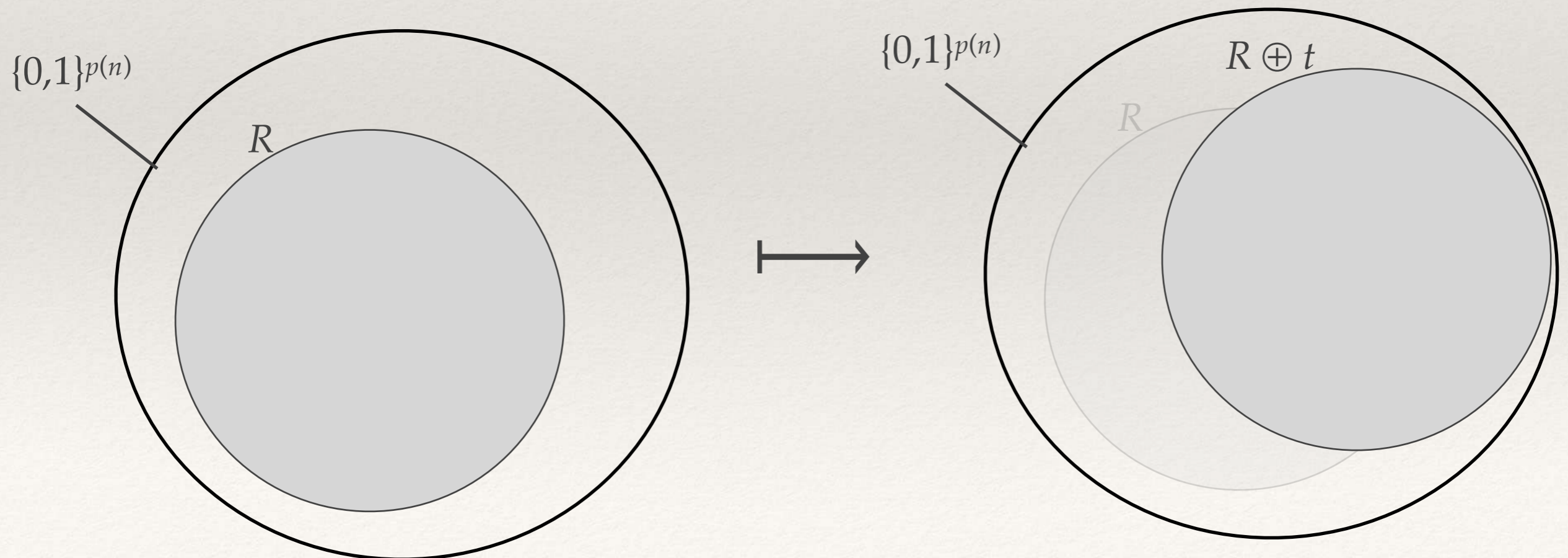
Translations?

- ❖ The algebraist's view: $\{0,1\}$ is the **field** $\mathbb{Z}/2\mathbb{Z}$,
 - exclusive or \oplus is **addition (mod 2)**
 - $\{0,1\}^{p(n)}$ is a $p(n)$ -dimensional **vector space**
 - and translation $R \oplus t = \{r \oplus t \mid r \in R\}$ is:



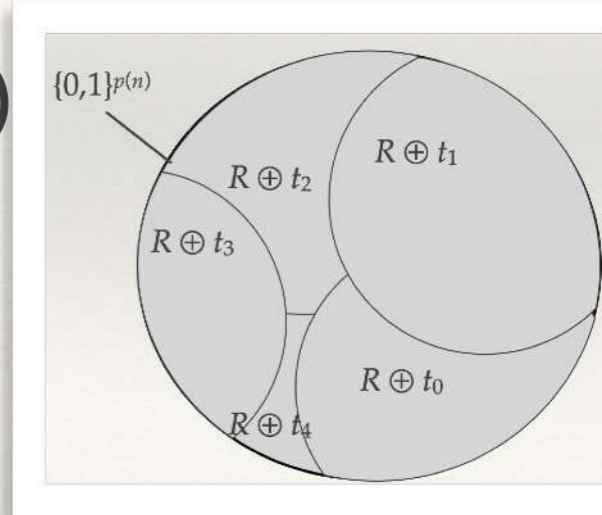
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The huge case (1/3)

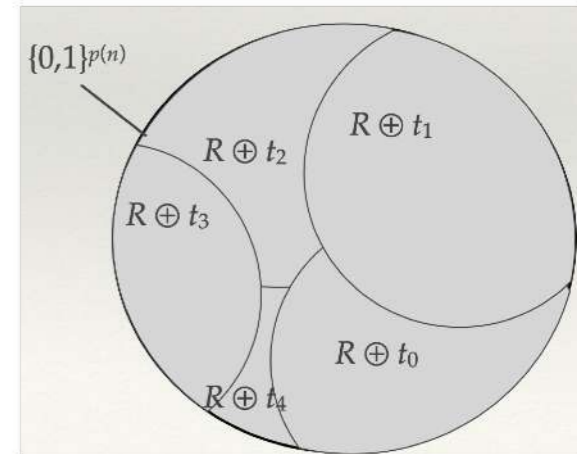
- ❖ Assume card $R \geq (1 - 1/2^n)2^{p(n)}$ (« R is huge »)
- ❖ **Claim.** $\exists t_0, \dots, t_{\lceil m/n \rceil}$ ($m=p(n)$) such that $R \oplus t_0, \dots, R \oplus t_{\lceil m/n \rceil}$ cover $\{0,1\}^m$.



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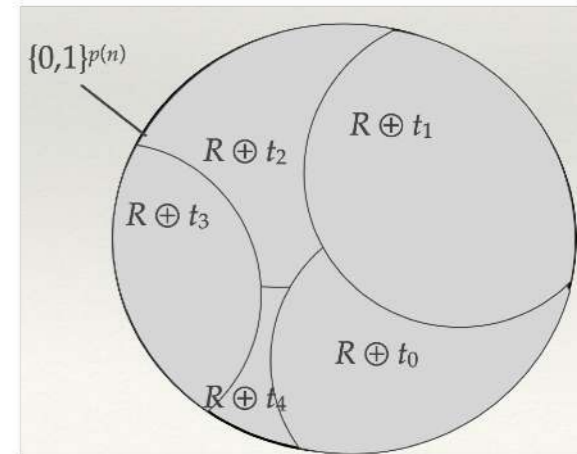


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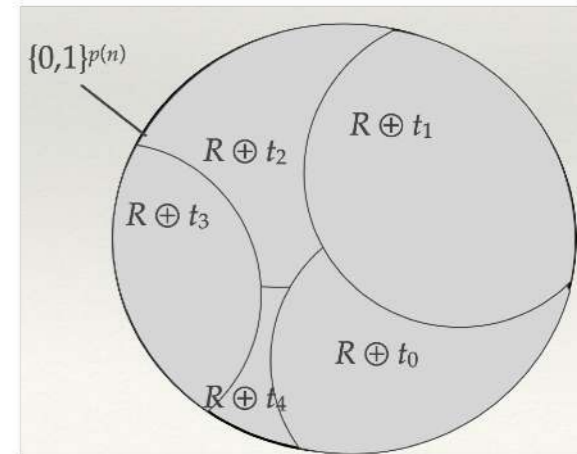
❖ By the probabilistic method. Let $\underline{t} = t_0, \dots, t_{\lceil m/n \rceil}$.



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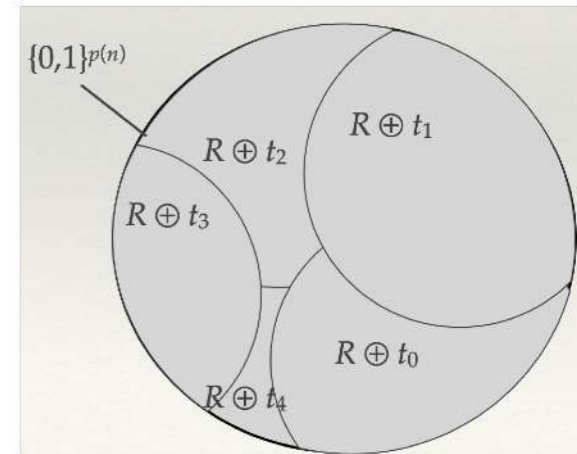
❖ By the **probabilistic method**. Let $\underline{t} = t_0, \dots, t_{\lceil m/n \rceil}$.

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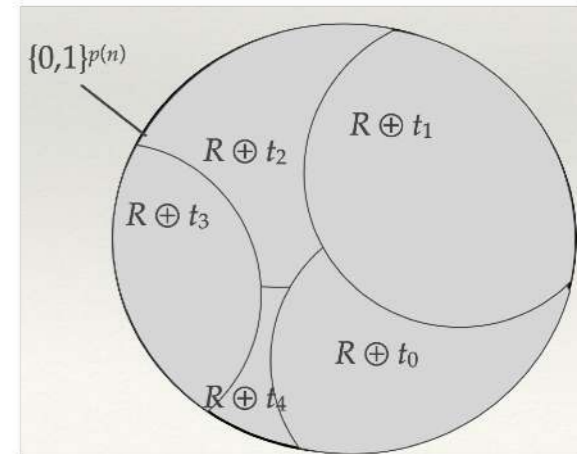
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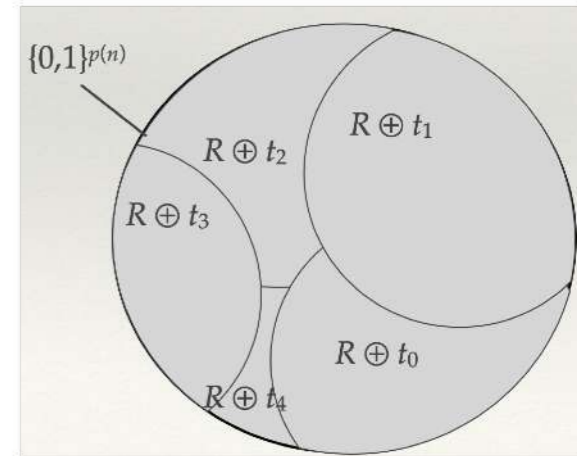
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Sum bound: $\Pr(\exists \dots) \leq \sum \Pr(\dots)$

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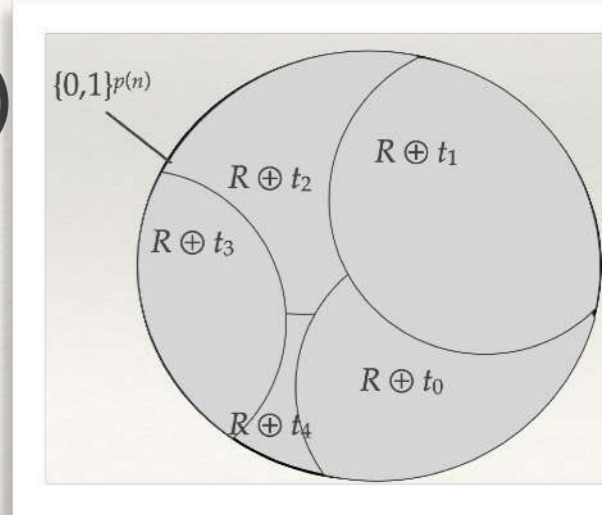
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Oh yes, that is a sum of $2^{p(n)}$ terms here!

The huge case (2/3)

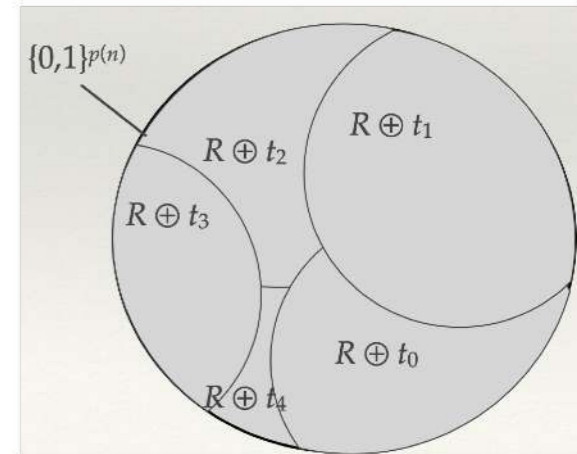
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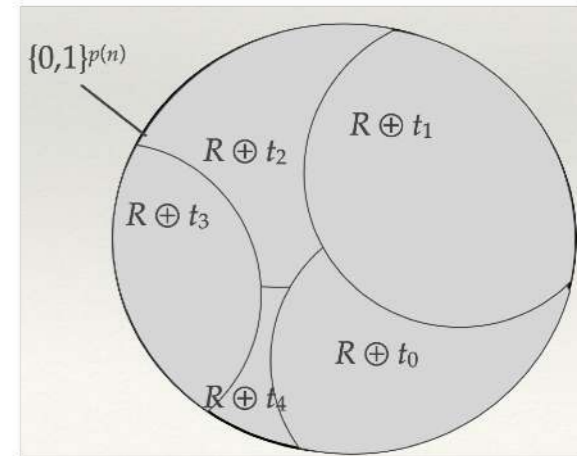
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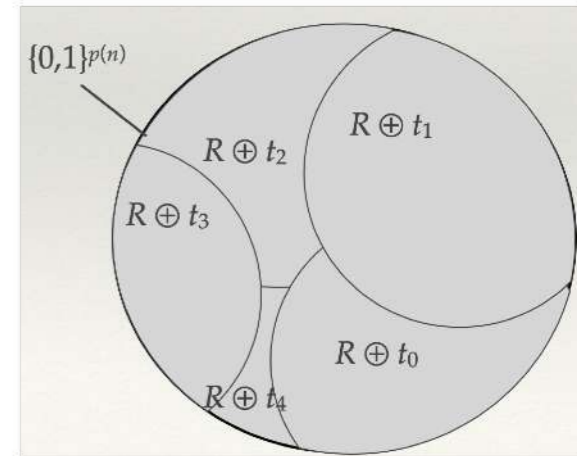
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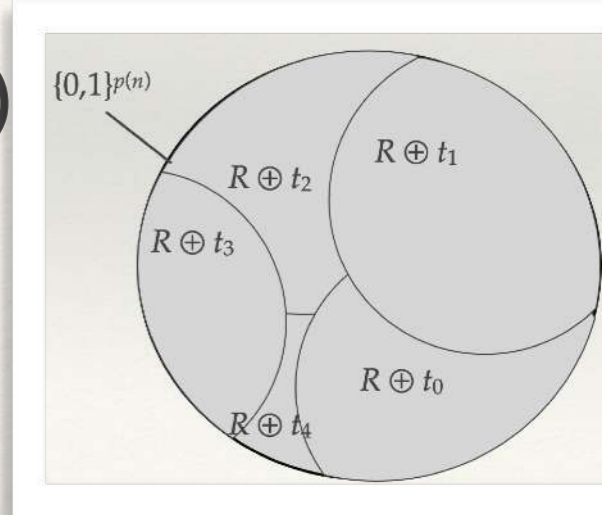
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($r \in R \oplus t$ iff $r \ominus t \in R$... but $\oplus = \ominus \pmod{2}$)

The huge case (3/3)

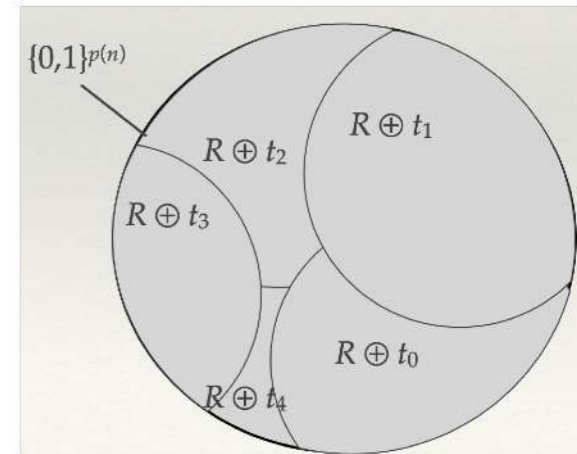
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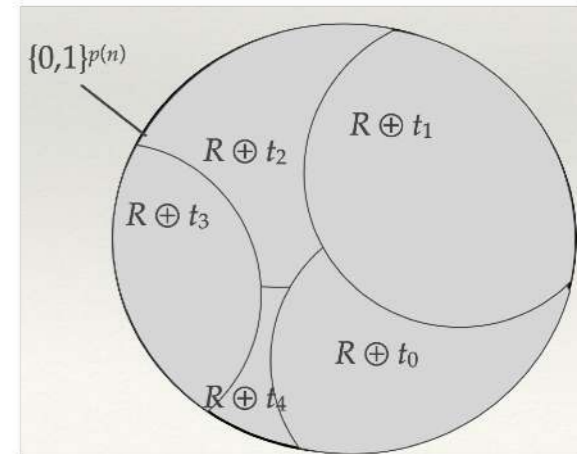
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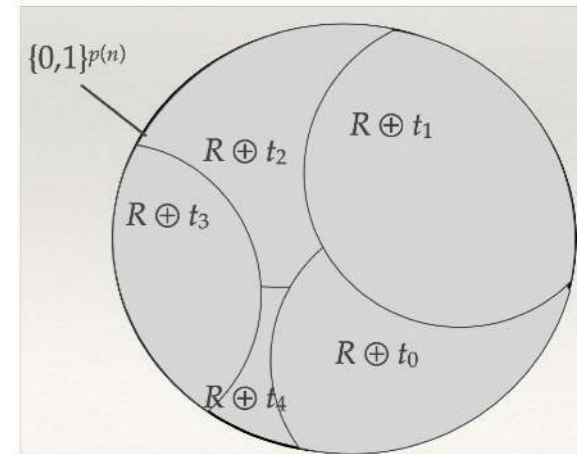
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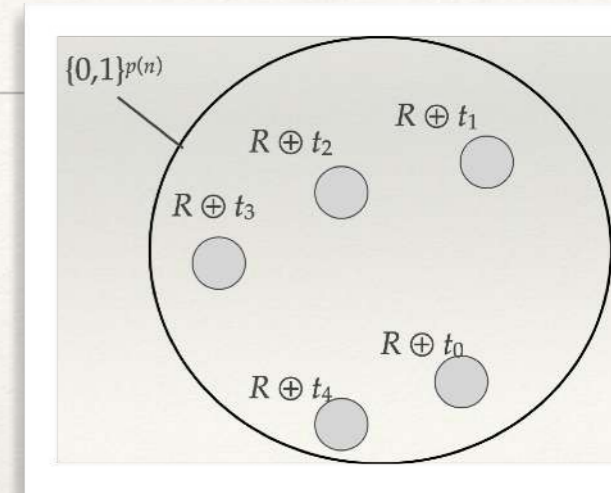
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❖ $\leq 2^m (1/2^n)^{\lceil m/n \rceil + 1} \leq 1/2^n < 1$ (at least if $n \neq 0$). Done! \square

The tiny case

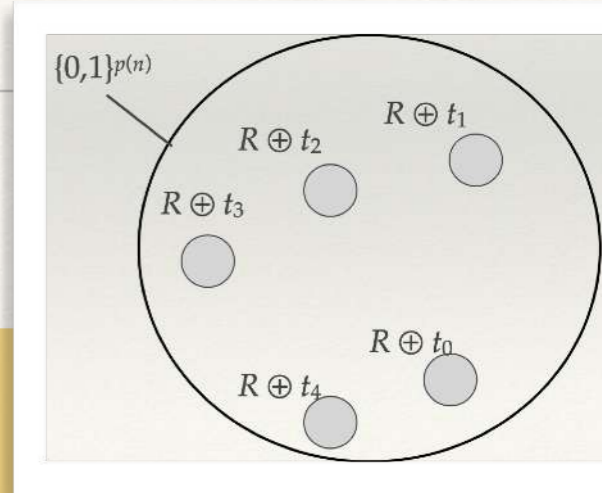
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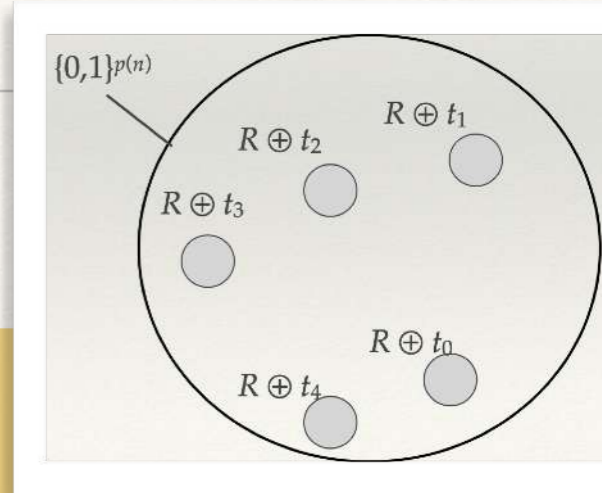


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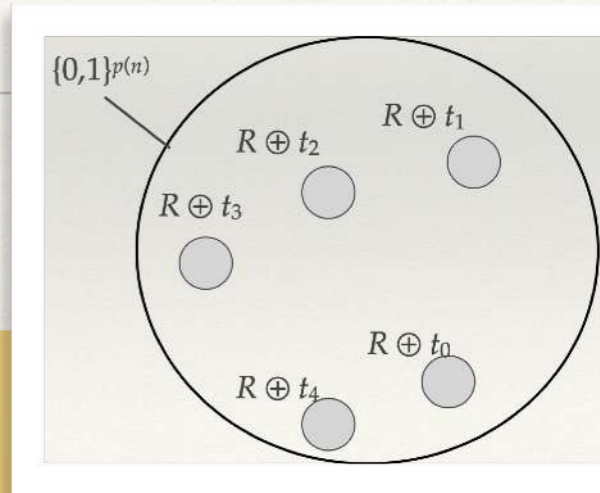
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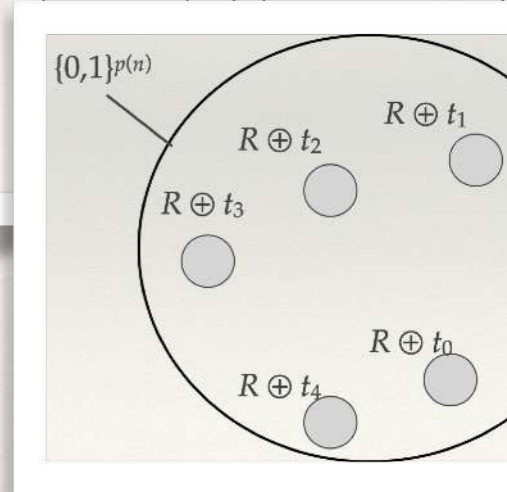
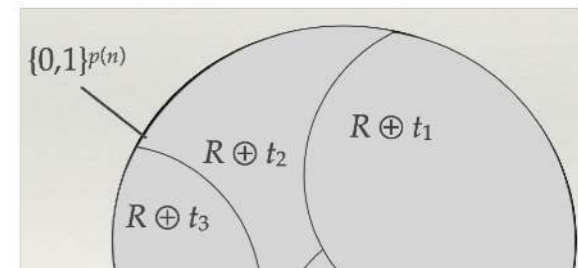
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❖ **strictly smaller** than card $\{0,1\}^m = 2^{p(n)}$
... if n large enough (say $n \geq n_0$). \square



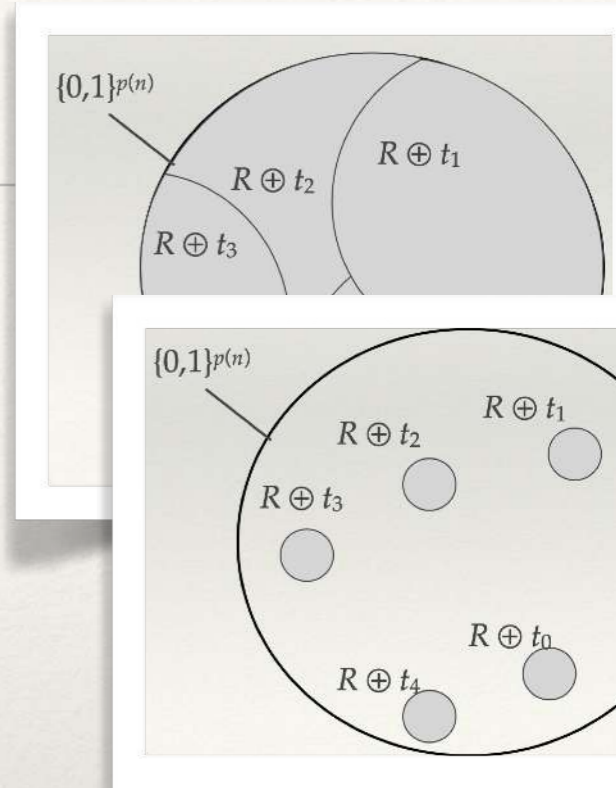
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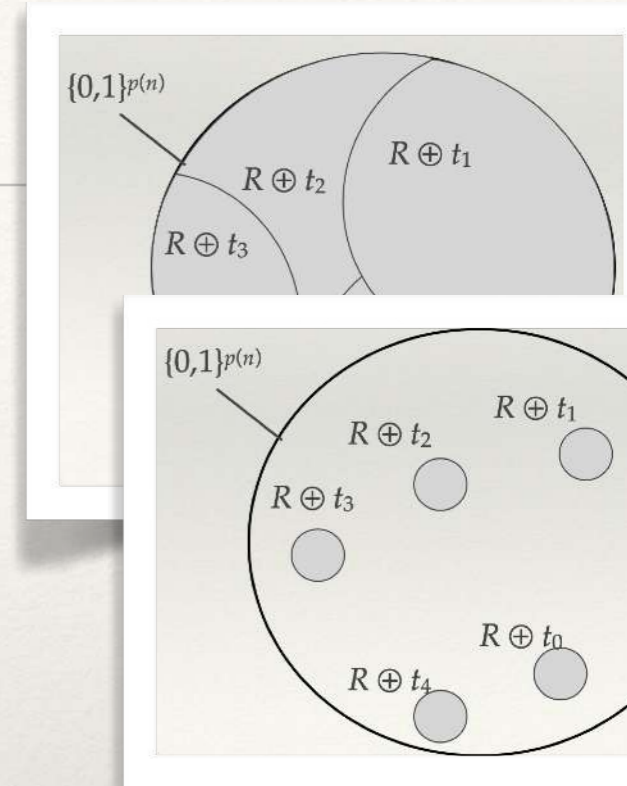
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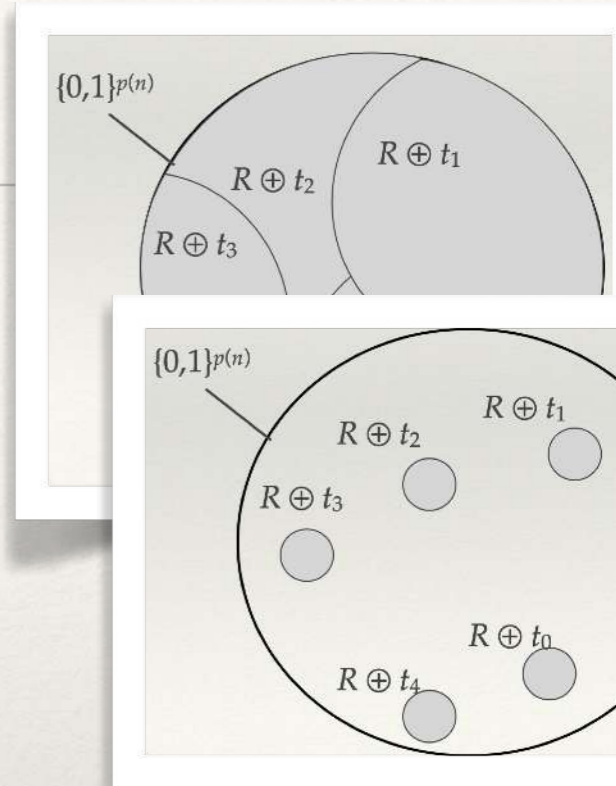
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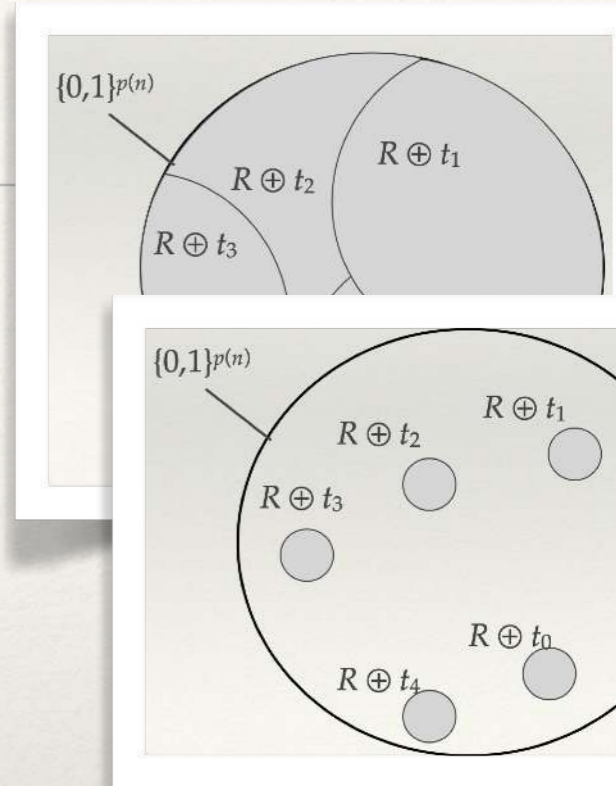
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- ❖ If $x \notin L,$ such $t_0, \dots, t_{\lceil m/n \rceil}$ do not exist (for $n \geq n_0$).



The algorithm

- ❖ Hence, for every x of size $n \geq n_0$,
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❖ Since L is arbitrary in **BPP**, $\mathbf{BPP} \subseteq \Sigma^P_2$. \square

The Sipser-Gács-Lautemann theorem

❖ **Theorem (Sipser-Gács-Lautemann, Prop. 1.24.)**

$$\mathbf{BPP} \subseteq \Sigma^{\text{P}_2} \cap \Pi^{\text{P}_2}.$$

❖ *End of proof.*

We have shown $\mathbf{BPP} \subseteq \Sigma^{\text{P}_2}$.

❖ Now $\mathbf{BPP} = \mathbf{coBPP} \subseteq \mathbf{co}\Sigma^{\text{P}_2} = \Pi^{\text{P}_2}$. \square

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Useful Lemma. Given two classes C_1, C_2 ,
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No, $\mathbf{co}C$ is **not** the complement of C .

It is the class of complements of languages in C .

Next time...

P/poly

- ❖ We will introduce a strange complexity class defined by **families of circuits**:
P/poly
- ❖ Studying it, we will eventually show that **BPP** probably does **not** contain **NP**
... otherwise **PH** would collapse at level 2!