Jean Goubault-Larrecq

Randomized complexity classes

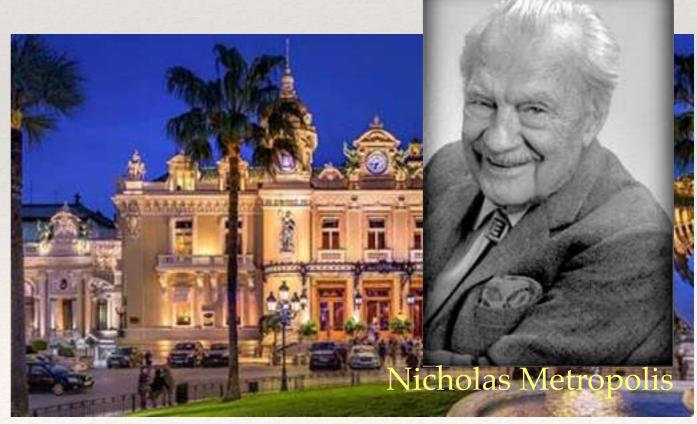
Today: BPP (part 1)

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)



http://fr.casino-jackpot.com/wp-content/uploads/2018/04/casino-monaco.jpg https://upload.wikimedia.org/wikipedia/commons/5/56/Nicholas Metropolis cropped.PNG

BPP: Bounded Prob. of Error Polynomial time

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

i.e. there is also a **polynomial** p(n) / $\mathcal{M}(x,r)$ terminates in time $\leq p(n)$, where n=|x|, in the worst case (and for any value of r)

... hence, implicitly, we require $|r| \ge p(n)$ (let us say |r| = p(n))

probability taken over all $r \in \{0,1\}^{p(n)}$

two-sided error:

 $\Pr_r\left[\mathcal{M}(x,r)\text{ errs}\right] \leq 1/3$

Examples



nago

discussion

view source

history

PolyMath

The complexity class BPP

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

Examples

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-HP, since it 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.

Error reduction

error = 1/3 here

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

- * Theorem. $\forall \epsilon \in]0, 1/2[$, $BPP = BPP(\epsilon)$.
- * Note: **BPP**=**BPP**(1/3) (def.)

BPP(ε)={all languages} if $\varepsilon \ge 1/2...$

 $\mathbf{BPP}(0) = \mathbf{P}$

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 $\Pr_r [\mathcal{M}(x,r) \text{ accepts}] \ge 2/3$ if $x \notin L$ then $\Pr_r [\mathcal{M}(x,r) \text{ accepts}] \le 1/3$.

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

The easy cases: error amplification(!)

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

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

The hard direction: repeating experiments

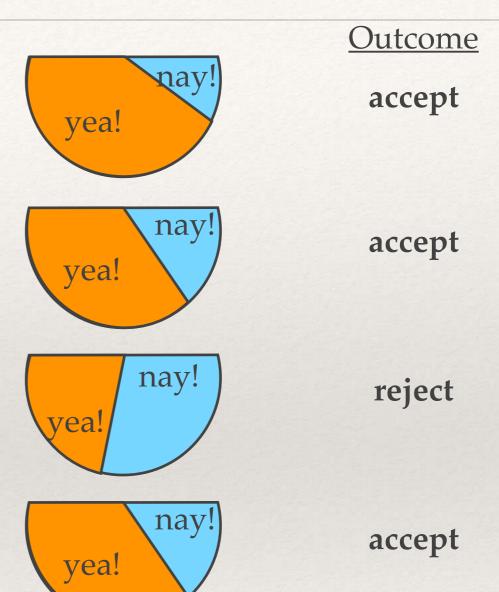
- * Let $L \in \mathbf{RP}(\varepsilon)$, $0 < \eta < \varepsilon < 1$
- On input x, let us do the following (at most) K times:
- * Draw r at random, simulate $\mathcal{M}(x, r)$ and:

- * A language L is in $\mathbb{RP}(\varepsilon)$ and only if there is a **polynomial-time** TM M such that for every input x (of size n):
- * if $x \in L$ then $\Pr_r [\mathcal{M}(x,r) \text{ accepts}] \ge 1-\varepsilon$
- * if $x \notin L$ then $\mathcal{M}(x,r)$ accepts for no ℓ (i.e., $\Pr_r[\mathcal{M}(x,r) \text{ accepts}] = \epsilon$
- * If $\mathcal{M}(x, r)$ accepts, then exit the loop and accept;
- Otherwise, proceed and loop.
- * At the end of the loop, reject.

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

Majority voting

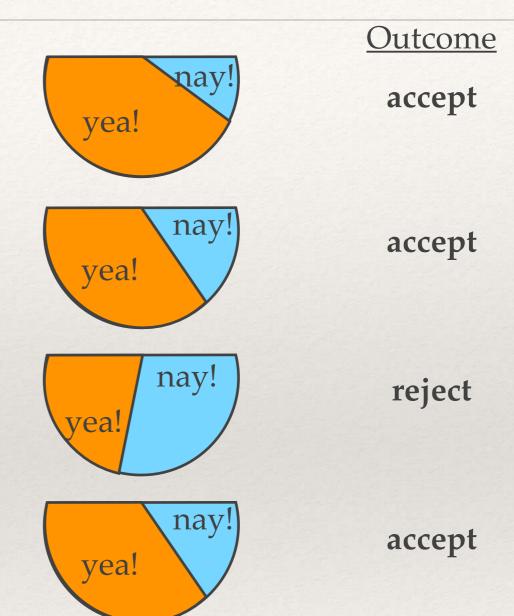
- Imagine running M(x,r) for various values of r, and tallying the votes
- * Redo the vote N times (here N=4)
- ♦ Here 3 accepts/1 reject
 ⇒ majority is for acceptance



Majority voting

* This is typical of what happens when $x \in L$: running a large number of votes should produce a majority of accepts, with high probability

* ... but how high?



Chernoff's bound

- * Intuitive contents:
 - Imagine Pr(yes) = p
 - Then Pr(proportion of **yes**es among N voters is close to p) goes to 1 **exponentially fast** as $N \rightarrow \infty$.
- * **Theorem.** Let $X_1, ..., X_N$ be **independent** rand. vars with values in $\{0, 1\}$ and

with the same law: $Pr(X_i=1)=p$.

Then $Pr(X_1+...+X_N\geq (1+\theta)pN)$

 $\leq \exp(-c(\theta)pN)$

We expect $X_1 + ... + X_N \approx pN$

 $1+\theta$ measures **how large** the deviation we allow for can be

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 $X = \exp(t(X_1 + ... + X_N))$

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



* Note that E(X) ≤ exp(tN) < ∞, so we can use **Markov's**

inequality:

 $\Pr(X \ge a.E(X)) \le 1/a$

Theorem (Markov's inequality).

Let X be a **non-negative real-valued** random variable with **finite** expectation E(X). For every $a \ge 0$: $Pr(X \ge a. E(X)) \le 1/a$.

Proof of Chernoff's bound (2/4)

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

*

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

$$\Pr(X \ge a.E(X)) \le 1/a$$

(from last slide)

- * Let us fix $a = \exp(t(1+\theta)pN) / E(X)$, hence:
- $Pr(X \ge \exp(t(1+\theta)pN)) \le E(X) \exp(-t(1+\theta)pN))$

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

Proof of Chernoff's bound (3/4)

- * Let t>0, to be fixed later
- $X = \exp(t(X_1 + \ldots + X_N))$

- Theorem. Let $X_1, ..., X_N$ be independent rand. vars with values in $\{0, 1\}$ and with the same law: $\Pr(X_i=1)=p$. Then $\Pr(X_1+...+X_N\geq (1+\theta)pN)$ $\leq \exp(-c(\theta)pN)$
- * $Pr(X_1+...+X_N\geq (1+\theta)pN) \leq E(X) \exp(-t(1+\theta)pN)$ (from last slide)
- * $E(X) = E(\Pi_{i=1}^{N} \exp(tX_{i}))$ $= \Pi_{i=1}^{N} E(\exp(tX_{i}))$ (independence) $= \Pi_{i=1}^{N} (p \exp(t) + 1 - p)$ (def. of the law of X_{i}) $= (p \exp(t) + 1 - p)^{N}$ $= (1 + p(\exp(t) - 1))^{N} \le \exp((\exp(t) - 1)pN)$

take logs: $N \log(1+p(\exp(t)-1)) \le Np(\exp(t)-1)$

Proof of Chernoff's bound (4/4)

- * Let t>0, to be fixed later
- $X = \exp(t(X_1 + \ldots + X_N))$
- * $Pr(X_1+...+X_N\geq (1+\theta)pN)$

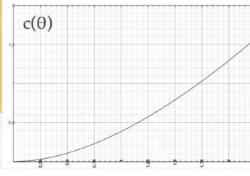
 $\leq \exp((\exp(t)-1)pN)\exp(-t(1+\theta)pN))$

(from last slide)

- * Let $t = \log(1+\theta)$, so $(\exp(t)-1)pN = \theta pN$, hence
- * $\Pr(X_1 + ... + X_N \ge (1+\theta)pN)$ $\le \exp((\theta - (1+\theta)\log(1+\theta))pN).$

Theorem. Let $X_1, ..., X_N$ be **independent** rand. vars with values in $\{0, 1\}$ and with the **same law**: $Pr(X_i=1)=p$.

Then $Pr(X_1+...+X_N \ge (1+\theta)pN)$ $\le exp(-c(\theta)pN)$

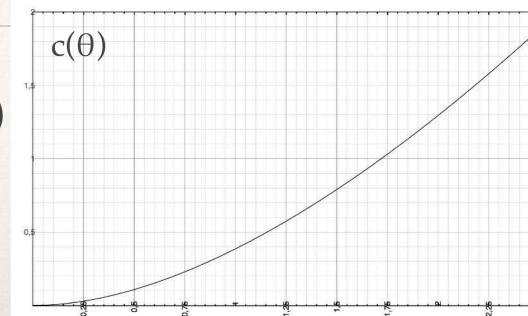


Call this $-c(\theta)$

Done!

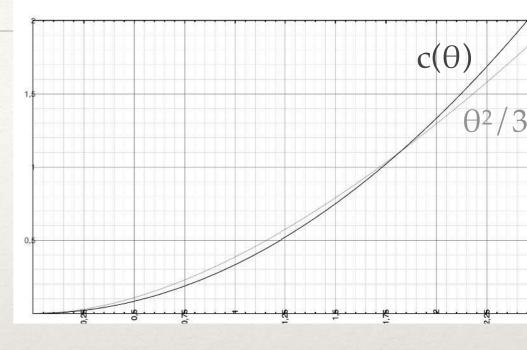
A few properties of $c(\theta) = -\theta + (1+\theta)\log(1+\theta)$

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



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

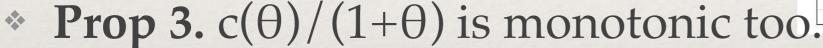
- * **Prop 1.** $c(\theta)$ is monotonic (for $\theta \ge 0$)
- * **Prop 2.** For $0 \le \theta \le 1$, $c(\theta) \ge \theta^2/3$
- * Proof. c(0)=0 c'(0)=0 (recall $c'(\theta) = \log(1+\theta)$) c''(0)=1 ($c''(\theta) = 1/(1+\theta)$) c'''(0)=-1 ($c'''(\theta) = -1/(1+\theta)^2$)

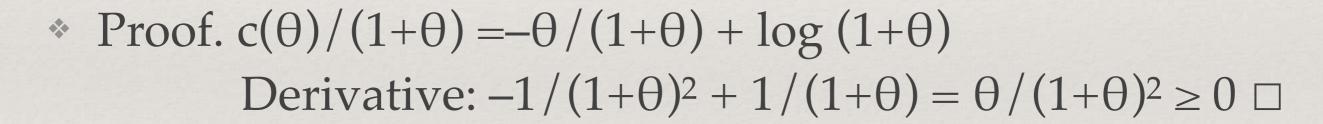


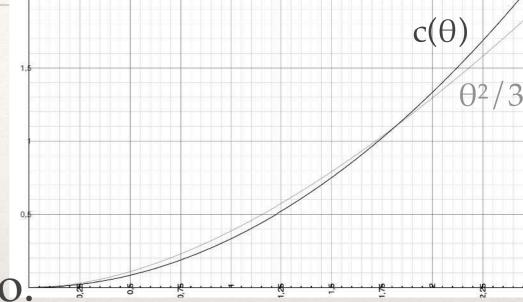
* So
$$c(\theta) = \frac{\theta^2}{2} - \frac{\theta^3}{6} + \frac{c^{(4)}(\theta_0)}{24}$$
 for some $0 \le \theta_0 \le \theta$ (Taylor) $\ge \frac{\theta^2}{2} - \frac{\theta^3}{6}$ (since $c^{(4)}(\theta) = \frac{2}{(1+\theta)^3} \ge 0$) $\ge \frac{\theta^2}{3}$ (since $\theta \le 1$)

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

- * **Prop 1.** $c(\theta)$ is monotonic (for $\theta \ge 0$)
- * **Prop 2.** For $0 \le \theta \le 1$, $c(\theta) \ge \theta^2/3$







Application to voting (1/4)

- * Assume that $\Pr_r(M(x,r) \text{ errs}) \le 1/3$, what is the probability P that more than 1/2 of N votes
 - $\mathcal{M}(x,r_1), \ldots, \mathcal{M}(x,r_N)$ err?
- * Let $X_i = 1$ iff $\mathcal{M}(x,r_i)$ errs: all assumptions satisfied with $p \le 1/3$

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Theorem. Let X_1, ..., 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+...+X_N\geq (1+\theta)pN) \leq \exp(-c(\theta)pN) (Chernoff)
```

* Take $\theta = 1/(2p)-1$, so $(1+\theta)p = 1/2$: $P \le \exp(-c(\theta)pN)$

Application to voting (2/4)

* 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 votes $\mathcal{M}(x,r_1)$, ..., $\mathcal{M}(x,r_N)$ err?

 $P \le \exp(-c(\theta)pN)$

Prop 2. For $0 \le \theta \le 1$, $c(\theta) \ge \theta^2/3$

- * Take $\theta = 1/(2p)-1$, so $(1+\theta)p = 1/2$:
- (from $\frac{1}{2}$ constant $\frac{1}{2}$ $\frac{1}{2}$
- * I.e., $P \le \exp(-c(\theta)/(1+\theta) \cdot 1/2 N)$
- * $\leq \exp(-c(1/2)/(3/2) \cdot 1/2 N)$ $\leq \exp(-c(1/2)/(3/2) \cdot 1/2 N)$ (since $p \leq 1/3$, so $\theta \geq 1/2$; plus Prop 3)
- * $\leq \exp(-(1/2)^2/3/(3/2) \cdot 1/2 N)$ (Prop 2) = $\exp(-N/36)$

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 votes $\mathcal{M}(x,r_1), ..., \mathcal{M}(x,r_N)$ err?
- * Answer: at most $\exp(-N/36)$

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

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

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

$$\Pr_r [\mathcal{M}(x,r) \text{ errs}] \leq \varepsilon.$$

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

accept if yeas $\ge N/2$, 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):

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

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

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

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

- * M' errs on input x iff at least half of the calls to M(x,r) err
- That happens with probability≤exp(-N/36)
- * ... $\leq \epsilon$ provided that we pick $N \geq -36 \log \epsilon$

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

- * Hence **BPP**(= **BPP**(1/3)) \subseteq **BPP**(ε) for ε arbitrarily close to 0
- * By a similar argument, we can replace 1/3 by any η , $0<\eta<1/2$, so $BPP(\eta)\subseteq BPP(\epsilon)$ for ϵ arbitrarily close to 0
- * Recalling that **BPP**(ε) \subseteq **BPP**(η) if $\leq \eta$, we obtain:
- * **Theorem.** For every ε , $0 < \varepsilon < 1/2$, **BPP=BPP**(ε).
- * ... 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 of N votes $M(x,r_1), ..., M(x,r_N)$ err

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 I $\mathcal{M}(x,r_1), \ldots, \mathcal{M}(x,r_N)$ err?
- * Answer: at most $\exp(-N/36)$

is
$$\leq 1/2^{q(n)}$$
?

- Answer: at least $36 q(n) \log 2$
- *Proof.* $\exp(-N/36) \le 1/2q^{(n)}$ iff $-N/36 \le -q(n) \log 2$

The only magical formula you'll need to remember for error reduction by majority voting

> Note: if q(n) is polynomial, this is polynomial, too

Error reduction for BPP revisited

- * Let L be in **BPP**
- * Build new rand. TM M' by:
- * yeas := 0 for i=1 to N := 36 q(n) log 2: draw r at random if $\mathcal{M}(x,r)$ accepts: yeas++

accept if yeas $\ge N/2$, else reject

- * M' errs on input x iff at least half of the calls to M(x,r) err
- That happens with probability

 $\leq 1/2q(n)$

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 \operatorname{BPP}(\varepsilon) nd 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-\varepsilon if x \notin L then \Pr_r[\mathcal{M}(x,r) \text{ accepts}] \leq \varepsilon
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error = \epsilon
```

- * **Theorem. BPP** is equal to:
 - **BPP**(ε) for every ε , $0 < \varepsilon < 1/2$
 - **BPP**(1/2q(n)) for every polynomial q(n)

The new landscape

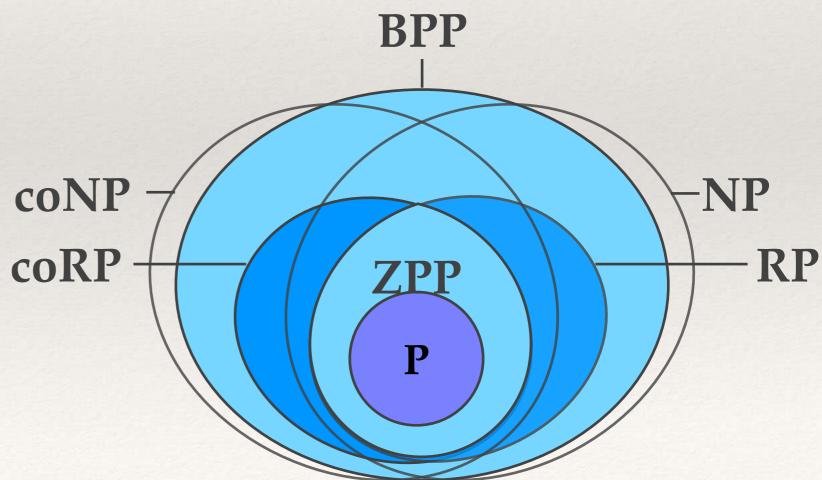
BPP vs. other complexity classes

* 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**(easy)

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



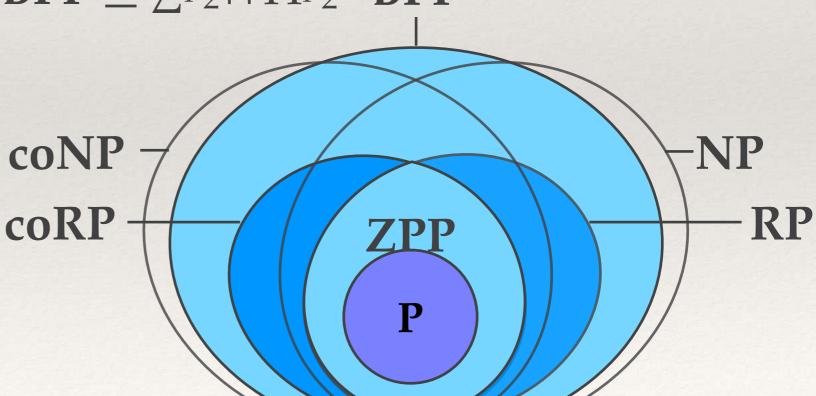
BPP cannot be too large

- * It is unknown whether $BPP \subseteq / \supseteq NP$ (eqv., coNP)
 - ... but we will see that $BPP \supseteq NP$ would have

drastic (and unlikely) consequences

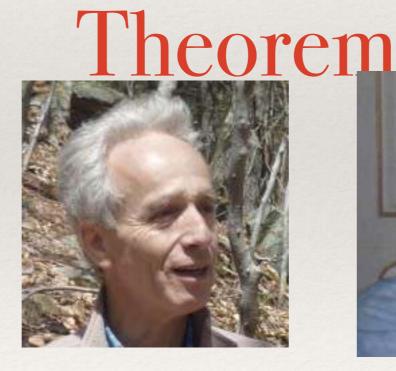
We start with this one

- * We will also see that $\mathbf{BPP} \subseteq \sum_{p_2} \cap \prod_{p_2} \mathbf{BPP}$
- * ... no significantly
 better result known! coNP
 although some coRP
 believe BPP=P.



The Sipser-Gács-Lautemann







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

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

The Sipser-Gács-Lautemann theorem

- * Theorem (Sipser-Gács-Lautemann, Prop. 1.24.) $BPP \subseteq \sum_{p_2} \cap \prod_{p_2} \prod_{p_$
- * Proof sketch.

It is enough to prove **BPP** $\subseteq \sum P_2$.

Proceeds by derandomization.

In order to do so, we will to prove

the existence of something

Funnily, this will involve Erdös' probabilistic method.

Of course: $\sum_{p_2} p_2$ is a non-randomized class...

$$\sum_{i=1}^{p_2} \mathbf{B} \cdot \mathbf{coNP}$$
$$(= \mathbf{A} \cdot \mathbf{V} \cdot \mathbf{P})$$

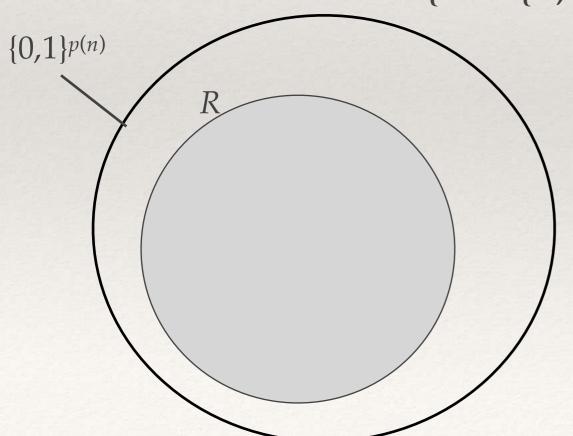
To prove that $\exists t, P(t)$, just show that $\Pr_t(P(t)) \neq 0$, or equivalently that $\Pr_t(\neg P(t)) < 1$

Lautemann's trick

* Let $L \in \mathbf{BPP}$, decided with error $\varepsilon = 1/2^n \pmod{1/3}$ in polytime p(n)

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

* Fix x. Then $R = \{r \in \{0,1\}^{p(n)} \mid \mathcal{M}(x,r) \text{ accepts} \}$ is



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

(covers a proportion $\geq (1-1/2^n)$ of the whole space)

error $\varepsilon = 1/2^n$

Lautemann's trick

* Let $L \in \mathbf{BPP}$, decided with error $\varepsilon = 1/2^n \pmod{1/3}$ in polytime p(n)

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

error $\varepsilon = 1/2^n$

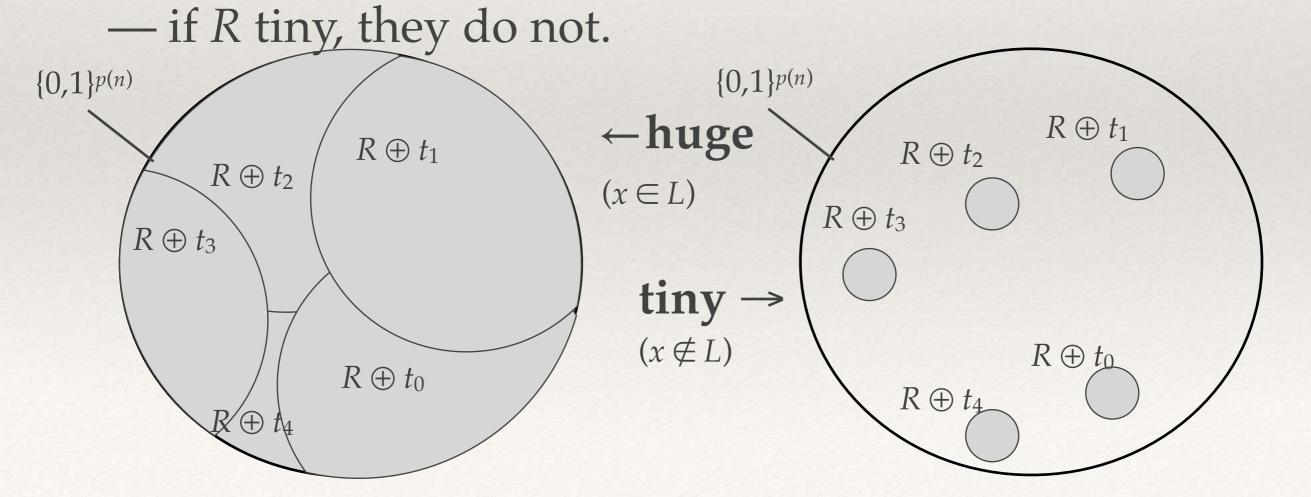
* Fix x. Then $R = \{r \in \{0,1\}^{p(n)} \mid \mathcal{M}(x,r) \text{ accepts} \}$ is

 $\{0,1\}p(n)$

or tiny, if $x \notin L$ (covers a proportion $\leq 1/2^n$ of the whole space)

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



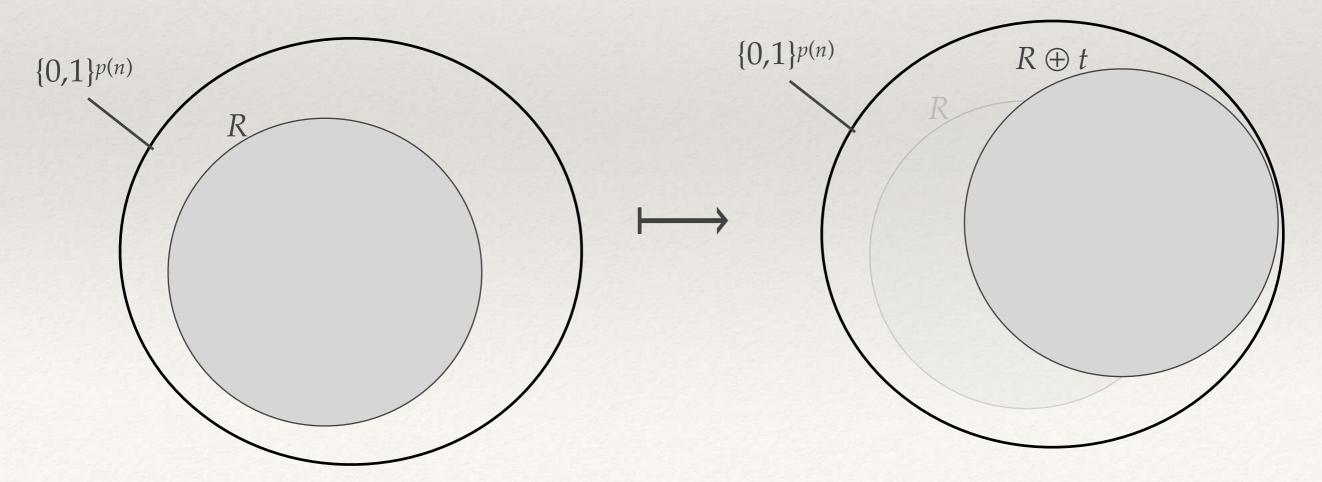
Translations?

- * The computer science view:
 - ⊕ is bitwise exclusive-or

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

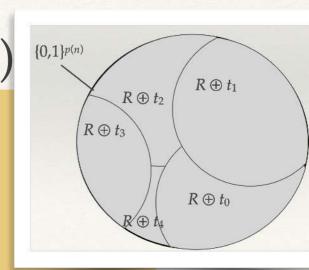
Translations?

- * 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:



The huge case (1/3)

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



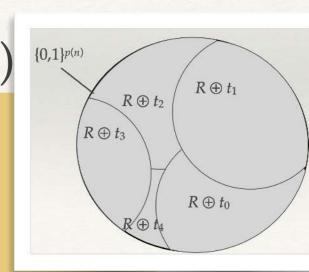
- * By the **probabilistic method**. Let $\underline{t}=t_0, ..., t_{\lceil m/n \rceil}$.
- * $\Pr_{\underline{t}}(R \oplus t_0, ..., R \oplus t_{\lceil m/n \rceil} \text{ does not cover } \{0,1\}^m)$
- $* = \Pr_{\underline{t}}(\exists r, r \notin R \oplus t_0 \text{ and } \dots \text{ and } r \notin R \oplus t_{\lceil m/n \rceil})$
- $* \leq \sum_{r} \Pr_{\underline{t}}(r \notin R \oplus t_0 \text{ and } \dots \text{ and } r \notin R \oplus t_{\lceil m/n \rceil})$

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

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

The huge case (2/3)

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



- * $\Pr_{\underline{t}}(R \oplus t_0, ..., R \oplus t_{\lceil m/n \rceil} \text{ does not cover } \{0,1\}^m)$
- * $\leq \sum_{r} \Pr_{\underline{t}}(r \notin R \oplus t_0 \text{ and } \dots \text{ and } r \notin R \oplus t_{\lceil m/n \rceil})$ (from last slide)
- $* = \sum_{r} \prod_{i=0}^{r} \Pr_{\underline{t}} (r \notin R \oplus t_i)$

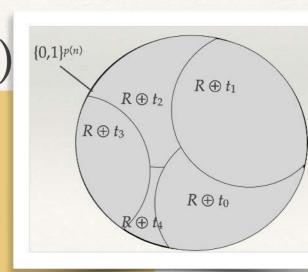
 $* = \sum_{r} \prod_{i=0}^{r} \Pr_{\underline{t}} (r \oplus t_i \notin R)$

(independence)

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

The huge case (3/3)

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



- * $\Pr_{\underline{t}}(R \oplus t_0, ..., R \oplus t_{\lceil m/n \rceil} \text{ does not cover } \{0,1\}^m)$
- * $\leq \sum_{r} \prod_{i=0}^{r} m/n^{r} \Pr_{\underline{t}} (r \oplus t_{i} \notin R)$ (from last slide)
- $* = \sum_{r} \prod_{i=0}^{r} \Pr_{t} (t \notin R)$

 $(t_i \mapsto t \stackrel{\text{\tiny def}}{=} r \oplus t_i \text{ bijection,}$ preserves cardinalities)

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

The tiny case

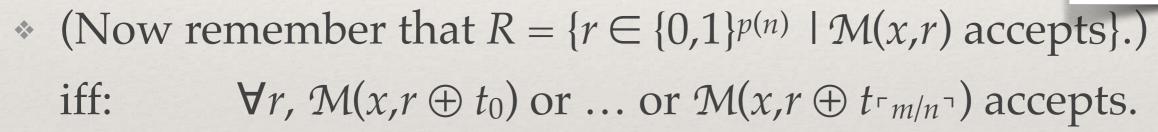
 $\{0,1\}p(n)$

if $n \ge n_0$

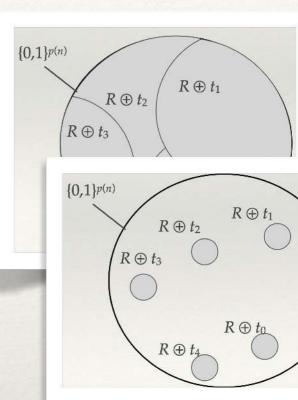
- * Assume card $R \le (1/2^n)2^{p(n)}$ (« R is tiny »)
- * Claim. $\forall t_0, ..., t_{\lceil m/n \rceil} \ (m=p(n)),$ $R \oplus t_0, ..., R \oplus t_{\lceil m/n \rceil} \ \text{does not cover } \{0,1\}^m.$
- * card $(\bigcup_{i=0}^{\lceil m/n \rceil} R \oplus t_i) \le (\lceil m/n \rceil + 1) (1/2^n) 2^{p(n)}$ = $O(\text{poly}(n)/2^n) 2^{p(n)}$
- * **strictly smaller** than card $\{0,1\}^m = 2^{p(n)}$... if n large enough (say $n \ge n_0$). \square

Testing huge vs. tiny

- * $R \oplus t_0, ..., R \oplus t_{\lceil m/n \rceil}$ covers $\{0,1\}^m$ iff: $\forall r, r \in R \oplus t_0 \text{ or } ... \text{ or } r \in R \oplus t_{\lceil m/n \rceil}$
- * iff: $\forall r, r \oplus t_0 \in R \text{ or } ... \text{ or } r \oplus t_{\lceil m/n \rceil} \in R$



- * If $x \in L$, $\exists t_0, ..., t_{\lceil m/n \rceil}$, $\forall r$, $\mathcal{M}(x,r \oplus t_0)$ or ... or $\mathcal{M}(x,r \oplus t_{\lceil m/n \rceil})$ accepts.
- * If $x \notin L$, such $t_0, ..., t_{\lceil m/n \rceil}$ do not exist (for $n \ge n_0$).



The algorithm

* Hence, for every x of size $n \ge n_0$, $x \in L$ iff $\exists t_0, ..., t_{\lceil m/n \rceil}$, $\forall r, \mathcal{M}(x,r \oplus t_0)$ or ... or $\mathcal{M}(x,r \oplus t_{\lceil m/n \rceil})$ accepts.

polytime (note $\lceil m/n \rceil = \lceil p(n)/n \rceil = poly(n)$)

- * For $n < n_0$, tabulate the answers.
- * Hence *L* is in $\sum P_2$.
- * Since *L* is arbitrary in **BPP**, **BPP** $\subseteq \Sigma^{p_2}$. \square

The Sipser-Gács-Lautemann theorem

- * Theorem (Sipser-Gács-Lautemann, Prop. 1.24.) BPP $\subseteq \sum_{p_2} \cap \prod_{p_2}$.
- * End of proof. We have shown **BPP** $\subseteq \Sigma^{p_2}$.
- * Now BPP = $\operatorname{coBPP} \subseteq \operatorname{co}_{\Sigma} p_2 = \prod p_2$. \square

Useful **Lemma.** Given two classes C_1 , C_2 , if $C_1 \subseteq C_2$ then $\mathbf{co} C_1 \subseteq \mathbf{co} C_2$. (Let $L \in \mathbf{co} C_1$. The complement of L is in C_1 hence in C_2 .)

No, **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!