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If **VV**  $\in$  P then SAT is in randomized poly time (RP).

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**When is a Randomized Algorithm Useful?** When it is fast and has a high probability of being correct.

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Is the Det of the above matrix 0? I do not know but I doubt it.

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**Note** In the above algorithm, we use “mod  $p$ ” so that the intermediate values do not get so large.



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$\text{Prob}(\text{DET}(M(a)) \equiv 0 \pmod{p}) = \text{Prob}(a \text{ is a root})$ :

$$\frac{dn}{d^2n^2} = \frac{1}{dn} \leq \frac{1}{n} \text{ which is small!}.$$

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**Moral** If have 1-sided error and Prob of error  $< 1$  then can iterate to get error very small.

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- 4) RP is thought to be feasible.



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**Plan** This small prob of success will get us all we need.

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# Our Plan (This is what Valiant-Vazirani did)

Given  $\phi$  we produce a formula  $\zeta$  such that

$\phi \in SAT \rightarrow \#(\zeta) = 1$  with high probability;

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# Hash Functions

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**Convention** Whenever we have a 0-1 valued matrix apply to a vector we do all of the calculations mod 2.

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**Note**  $E(S)$  and  $\text{Var}(S)$  do not depend on  $n$ , just on  $k$  and  $|X|$ .

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$$E(R_x R_y) = \Pr(M(x) = 1 \wedge M(y) = 1) = \frac{1}{2^k} \times \frac{1}{2^k} = \frac{1}{4^k}.$$

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The matrix question is easier: By convention the  $0 \times n$  matrix has no effect. So

$$X = \{x \in X : M(x) = 0^k\}.$$

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- 1)  $\text{SAT} \leq_r \text{SAT}_{12}$ . (Why 12? We'll see later.)
- 2)  $\text{SAT}_{12} \leq_r \text{SAT}_1$ . (Not Quite- this reduction will only be correct if the input comes from the first reduction.)

$$\text{SAT} \leq_r \text{SAT}_{12}$$

# Chebyshev's inequality

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Chebyshev proved it so we don't have to :-)

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**That is all we need to show!**

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$SAT_{12} \leq_r SAT_1$   
**Not Quite**

# What We Really Need

**Recall** that we have a reduction that maps  $\phi$  to  $\psi$  such that

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We will get (with restricted input)

$$\begin{aligned}\psi \in \text{SAT}_{12} &\rightarrow \Pr(\zeta \in \text{SAT}_1) \geq \frac{1}{12} \\ \psi \notin \text{SAT} &\rightarrow \zeta \notin \text{SAT} \text{ hence } \zeta \notin \text{SAT}_1\end{aligned}$$

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We need  $\text{SAT}_{12} \leq_r \text{SAT}_1$  where the input  $\psi$  has

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We will get (with restricted input)

$$\begin{aligned}\psi \in \text{SAT}_{12} &\rightarrow \Pr(\zeta \in \text{SAT}_1) \geq \frac{1}{12} \\ \psi \notin \text{SAT} &\rightarrow \zeta \notin \text{SAT} \text{ hence } \zeta \notin \text{SAT}_1\end{aligned}$$

Compose the two prob reductions to get  $\text{SAT} \leq_r \text{SAT}_1$ .

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**Analysis** on next slide.

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We are done!

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- 5) By Lemma, if  $\text{SAT}_1 \in \text{P}$  then  $\text{SAT} \in \text{RP}$ .
- 6) One can modify to get: if  $\text{SAT}_1 \in \text{RP}$  then  $\text{SAT} \in \text{RP}$ .

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