

- Computational Pseudorandomness.
- Blum-Micali-Yao paradigm: Based on 1-way functions.
- Nisan-Wigderson paradigm: Based on hard functions.

sequence is not random, else it is. What about finite sequences?

- Blum-Micali-Yao: Random is when can't predict the next bit in time polynomial in  $n$ . Equivalently, set is pseudorandom if no polynomial time algorithm behaves differently on string than on uniform distribution.

- Initially: Sequence of random bits, independent, uniform are random. Nothing else is.
- Shannon: Uniformity not necessary. Independence not necessary. Can attribute to any distribution an entropy which measures amount of randomness in it. Roughly - uniform distribution on  $S \subseteq \{0, 1\}^n$  has  $\log_2 |S|$  bits of entropy. "Large sets are be random; small sets aren't".
- Kolmogorov-Chaitin-Solomonoff: "Random is what can't be described": If finite TM produces an infinite sequence, then

### BMV: Strings vs. Sets of Strings

- Roughly: A fixed finite length string can't be random in any meaningful sense.
- However a long string generated from a short random seed can appear random to some.
- What is random?
  - If you can't distinguish given distribution from random, then distribution is pseudo-random to you.
  - if  $You = \{ \text{Class of all polytime algorithms (circuits)} \}$ , then distribution is pseudo-random.

## BMV: Indistinguishability as random

- Distributions  $D_1$  and  $D_2$  are  $\epsilon$ -indistinguishable to Boolean  $A$  if

$$|\Pr_{x \leftarrow D_1}[A(x) = 1] - \Pr_{x \leftarrow D_2}[A(x) = 1]| \leq \epsilon.$$

- Distributions  $D_1$  and  $D_2$  are statistically indistinguishable if they are  $1/p(n)$  indistinguishable to every  $A$  and every polynomial  $p$ .
- Distributions  $D_1$  and  $D_2$  are computationally indistinguishable if they are  $1/p(n)$  indistinguishable to every polytime computable function (poly size circuit)  $A$  and every polynomial  $p$ .

- BMV Notion:  $D_1$  is pseudo-random if it is computationally indistinguishable from uniform distribution.

## BMV: Pseudorandom generators

- $G : \{0, 1\}^s \rightarrow \{0, 1\}^n$  is a pseudorandom-generator if  $\{G(s)\}_s$  is computationally pseudo-random and  $G$  is polytime computable in input.
- Note  $G$  is easy, but  $G^{-1}$  hard. Thus prg needs  $NP \neq P$ . Even more!  $DNP \neq Avg - P$ .
- Focus on polynomial length stretching, not more.

## BMV: Alternately, Unpredictable is random

- $i$ th bit of  $G$  is  $\delta$ -unpredictable to  $A$  if  $\Pr_s[G(s)[i+1] = A(G(s)[1, \dots, i])] \leq \frac{1}{2} + \delta$ .
- $G$  pseudorandom if for all  $i$  and for all prob. poly time  $A$ , and all  $\delta = 1/p(n)$ ,  $i$ th bit of  $G$  is  $\delta$ -unpredictable to  $A$ .
- Thm: Two defs are equivalent.
- Proof. One direction obvious. Other direction is hybridization + case analysis.

## Consequence: 1-bit stretcher suffices

- Let  $G$  map  $s$  bits to  $s + 1$ .
- Will construct prg mapping  $s$  bits to  $n$  from this.
- Let  $S_0 = S$  be initial seed. Let  $x_i = G(S_{i-1})$  and let  $S_i =$  first  $s$  bits of  $x_i$  and let  $y_i =$  last bit of  $x_i$ . Then the map from  $S$  to  $y_1 \cdots y_n$  is pseudorandom.
- Proof: Consider a predictor predicting  $y_i$  given  $y_{i+1} \cdots y_n$  (Aha! Reversing the output). Then the predictor can also predict  $y_i$  given  $S_i$  (since  $y_{i+1} \cdots y_n$  can be computed from  $S_i$ ). But this is predicting the last bit in the  $i$ th application of  $G$ !

## Constructions & Applications

- First notice we didn't really need  $G$  to be pseudo-random, only that its last bit be unpredictable given the first  $s$ .
- Blum-Micali: Prove that the map  $G : (p, g, x) \mapsto (p, g, g^x \pmod{p}, \text{msb}(x))$  satisfies this property if we believe discrete log. to be hard. Use this to construct prg.
- Applications: Mostly in cryptography. Often easy to show that "knowledge" is not leaked by some string, by showing it is computationally pseudorandom.
- Our quest: Complexity-theoretic use. Use pseudo-randomness to show  $\text{BPP}=\text{P}$ . First steps by Yao. Later Nisan-Wigderson.

## Nisan-Wigderson paradigm

- $G : \{0, 1\}^s \rightarrow \{0, 1\}^n$  is a pseudorandom-generator if  $\{G(s)\}_s$  is computationally pseudo-random to circuits of size  $n$  and  $G$  is polytime computable in output.
- Now don't need to show  $\text{NP} \neq \text{P}$ ! Still need to show some function in, say,  $\text{time}(n^2)$  does not have size  $n$  circuits.
- Main theorem: Suffices to have such functions.
  - Step 1: If function in  $E$  is hard on average for subexp. circuits then  $\text{BPP}=\text{P}$ .
  - Step 2: If function in  $E$  is hard on worst-case for subexp. circuits then there exists function in  $E$  is hard on average.