Interactive Proofs

Or how I stopped worrying and learned to ask questions

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• If there is no proof, then the *given statement* must be false

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- Here, the verifier V, is a polynomial time Turing machine which takes strings of a language L and outputs 1 if the string is in L or 0 otherwise.

• The Prover P, is a <u>function</u> that maps strings to a certificate or "Sorry, not in the language".

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• Provers are always trying to prove correctness, even if a statement is not correct.

• Even if the prover diligently says that there is no proof, the verifier cannot be sure unless the verifier knows that the prover is **all powerful**.

Interactive Proof systems: The Protocol

Definition: Let f, g : $\{0, 1\}^* \rightarrow \{0, 1\}^*$ be functions. A k-round interaction of f and g on input $x \in \{0, 1\}^*$, denoted by <f, g>(x) is the sequence of the following strings $a_1, \ldots, a_k \in \{0, 1\}^*$ defined as follows:

 $a_1 = f(x)$ $a_2 = g(x, a_1)$...

$$a_{2i+1} = g(x, a_1, \dots, a_{2i+1})$$

 $a_{2i+2} = g(x, a_1, \dots, a_{2i+1})$

 $a_{1} = f(x_1, a_2, \dots, a_n)$

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Definition: Deterministic proof systems

For $k \ge 1$, We say that a language L has a k-round deterministic interactive proof system if there's a deterministic poly-time TM V that on input x, a_1, \ldots, a_i runs in time polynomial in |x|, satisfying:

 $x \in L \Rightarrow \exists P : \{0, 1\}^* \rightarrow \{0, 1\}^* \text{ out}_{\vee} < V, P > (x) = 1 \text{ (Completeness)}$ x ∉ L ⇒ ∀P : {0, 1}* → {0, 1}* out_{\u03c0} < V, P > (x) = 0 (Soundness)



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• **dIP** is the set of all languages with poly(n)-round deterministic interactive proof system.

• Can't we define a class of constant round deterministic interactive proof systems?



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Proof: One round protocol for 3SAT, where a prover returns a satisfying assignment for the input if it exists.



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• Lemma: dIP = NP



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IP: Probabilistic Verifier

Definition [GMR '89]: Probabilistic Verifiers and IP

For $k \ge 1$, we say that a language L has in **IPTIME**[k] if there's a probabilistic poly-time TM V that has a k-round interaction with P: $\{0,1\}^* \rightarrow \{0,1\}^*$ that on input x

$$x \in L \Rightarrow \exists P Pr_r[out_V < V, P > (x) = 1] \ge 2/3$$
 (Completeness)

 $x \notin L \Rightarrow \forall P Pr_{r}[out_{\vee} < V, P > (x) = 1] \le 1/3$ (Soundness)

The probabilities over the random bits r of V.

The class IP is defined as $IP = U_{i>0}$ $IPTIME[n^c]$

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Proof: Similar to boosting a BPP machine. Polynomially(n^c) many independent repetitions of protocol.

Additionally, we can also do all repetitions in parallel by asking multiple questions in each round, thereby decreasing the number of rounds.



What's in IP?

• Clearly, NP is also in IP.

As dIP is in IP

• So is BPP

The verifier is a BPP machine that ignores the prover



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NONISO = { (G_1, G_2) | G_1 is not isomorphic to G_2 }

• Lemma: NONISO \in IP [GMW '91]

Private Coin Protocol

1. V randomly picks a graph between G_1 and G_2 , say G_i . Randomly permute vertices of G_i to make H. Send H to P and asks if H is isomorphic to G_1 or G_2

2. Prover tries to figure out whether H is isomorphic to G_1 or G_2 , sends $j \in \{1,2\}$ to V

3. V accepts if j==i.





V



V














V



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• If G_1 and G_2 are not isomorphic, then the prover should be able to compare every permutation of H with G_1 and G_2 to be able to answer correctly.

• The probability of acceptance when the string is in the language is 1. (Perfect Completeness)

 If they are not isomorphic, the best the prover can do is to guess at random. So the probability of acceptance when it isn't in the language is ¹/₂.
We can decrease this be multiple repetitions.

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What about random bits of V?

Constant round interactive proofs with public coins: AM and MA.

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• Theorem [Babai '88]: AM[k] = AM[2] for constant k

• Theorem [GS '86]: $AM[k] \subseteq IP[k] \subseteq AM[k+2]$ for polynomial k.

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Proof: Consider a language A in IP with a verifier V. Let the transcript be exactly of size p = poly(n) for all inputs x of size n. We will construct a PSPACE machine M which decides A.

Theorem: IP ⊆ PSPACE

Definition: For any string x, we define

 $Pr[V accepts x] = max_{P} Pr[\langle V, P \rangle accepts x]$

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Definition: $\langle V, P \rangle (x, r, M_j)$ = accept, for a random string r of length p, if there exists $m_{j+1}, ..., m_p$ such that

- 1. For $j \le i < p$ and i is even $V(x,r,M_i) = m_{i+1}$
- 2. For $j \le i < p$ and i is odd $P(x, M_i) = m_{i+1}$
- 3. m_p is accept

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Obs: Using previous definitions,

 $Pr[\langle V,P \rangle accepts x starting at M_{j}] = Pr[\langle V,P \rangle (x,r,M_{j}) = accept]$ (1)

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The goal is now to compute the probability of V accepting x starting from M_0 . If this is greater than $\frac{2}{3}$ then x must be in A, if it less than $\frac{1}{3}$ then it must not be in A. We do this recursively.

Theorem: IP ⊆ PSPACE

 $N_{M_j} = 0$ if j = p and $m_p = reject$ = 1 if j = p and $m_p = accept$ $= \max_{m_{j+1}} N_{M_{j+1}}$ odd j < p $= wt\text{-}avg_{m_{j+1}} N_{M_{j+1}}$ even j<p wt-avg_{m_{j+1}} $N_{M_{j+1}} = \sum_{m_{j+1}} ((Pr[V(w,r,M_j)=m_{j+1}]) \cdot N_{M_{j+1}})$

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Base case: j = p. The last message must be *accept* or *reject*. Hence, the probability of acceptance when the last message is *reject* is 0 and when the last message is *accept*, it is 1. This is exactly how N_{M_i} is defined.

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Inductive step: Assume the claim to be true for some $j+1 \le p$. We have 2 cases, one when j is even and when j is odd.

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When j is odd, the message m_{j+1} is from P to V. From the definition of N_{M_j}

 $N_{\text{M_j}} = max_{m_\{j+1\}} N_{\text{M}_\{j+1\}}$

 $N_{M_{j+1}} = \max_{m_{j+1}} \Pr[V \text{ accepts } x \text{ starting at } M_{j+1}]$

$$= \max_{m \{i+1\}} \max_{P'} \Pr[\langle V, P' \rangle (x, r, M_{i+1}) = accept]$$

 $1... \le \max_{P} \Pr[\langle V, P \rangle \text{ accepts x starting at } M_i], P can send the maximizing m^*_{i+1}$

 $2... \ge \max_{P} Pr[\langle V, P \rangle accepts x starting at M_i], P cannot be better than P'$

Therefore,

 $N_{M_j} = Pr[V \text{ accepts } x \text{ starting at } M_j]$

Claim 2: N_{M_i} can be calculated in PSPACE

From the above proof, it also clear that these values can be calculated in PSPACE recursively. The depth of the recursion would be p. M calculates $N_{M_{\perp}}$ for every j and M_{i} .

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 We can prove P^{#P} ⊆ IP if #3SAT is in IP, would automatically imply PH is in IP by Toda's theorem

• Proven by [LFKN '92]

#3SAT Prerequisites

• Definition: #3SAT

#3SAT = {(ϕ ,k)| where ϕ is a 3CNF with exactly k satisfying assignments}

 $\#\phi$ is the number of satisfying assignments of 3CNF ϕ

Say $\phi(x_1, \dots, x_n)$, then

$$\#\phi = \sum_{b_1 \in \{0,1\}} \sum_{b_2 \in \{0,1\}} \dots \sum_{b_n \in \{0,1\}} \phi(b_1, \dots b_n)$$

 $\phi(b_1,...,b_n) = 1$ if $b_1...,b_n$ is a satisfying assignment, 0 otherwise

We define $\#\phi(a_1,...a_{i-1})$ as

$$\#\phi(a_1,...,a_{i-1}) = \sum_{b_i \in \{0,1\}} ... \sum_{b_n \in \{0,1\}} \phi(a_1,...,a_{i-1},b_i,...,b_n)$$

#3SAT Prerequisites

Observation*: $\#\phi(a_1,...,a_{i-1}) = \#\phi(a_1,...,a_{i-1},0) + \#\phi(a_1,...,a_{i-1},1)$

$$\#\phi(a_1,...,a_{i-1}) = \sum_{b_i \in \{0,1\}} ... \sum_{b_n \in \{0,1\}} \phi(a_1,...,a_{i-1},b_i,...,b_n)$$

$$= \Sigma_{b_{\{i+1\} \in \{0,1\}}} ... \Sigma_{b_{n} \in \{0,1\}} \phi(a_{1}, ..., a_{i-1}, 0, ..., b_{n}) + \Sigma_{b_{\{i+1\} \in \{0,1\}}} ... \Sigma_{b_{n} \in \{0,1\}} \phi(a_{1}, ..., a_{i-1}, 1, ..., b_{n})$$

$$= \#\phi(a_1, \dots, a_{i-1}, 0) + \#\phi(a_1, \dots, a_{i-1}, 1)$$

Say the input is (ϕ ,K). The verifier has to check whether ϕ indeed has K satisfying assignments. Try to verify observation*

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- 6. Repeat by setting each variable x_i to 0 and 1 and verifying
- 7. Step ??: Once all variables have been set, Verifier asks the prover the number of satisfying assignments and also verifies the answer by itself.
\$\$(X₁,...**X**_n**)**

φ(x₁,..x_n) **K**



























φ(0,0,..0) <u>1</u>



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φ(0,0,..0)=1 == 1



 $\phi(0,0,..0)$ <u>1</u> $\phi(0,1,..1)$ <u>0</u>

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Idea: Randomly choose a path in the tree

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φ(x₁,..x_n) K


















. . .



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φ(1,0,..1) ¹

φ(1,0,..1)=0 != 1

#3SAT \subseteq IP? [Attempt 2]

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• The probability that the prover actually gets caught is 2⁻ⁿ. We need to catch every wrong branch at every step.

 So, we always accept when the number of satisfying assignments are correct, but we will also accept when it is incorrect with probability 1 - 2⁻ⁿ.

Every boolean formula can be expressed as a polynomial over elements of F_2

We use the following trick:

 $a \wedge b \equiv ab$ $a \vee b \equiv 1 - (1-a)(1-b) \equiv a + b - ab$ $\neg a \equiv (1-a)$ True = 1 False = 0

Example:

$$(x_1 \vee x_3 \vee \neg x_4) \equiv (x_1 + x_3 - x_1 x_3) + (1 - x_4) - (x_1 + x_3 - x_1 x_3)(1 - x_4)$$

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 The size of the polynomial will also be bound polynomial in the size of φ as we don't need to expand the terms

We can restate our equations as follows, where X_is are now formal variables

$$#\phi = \sum_{b_1 \in \{0,1\}} \sum_{b_2 \in \{0,1\}} \dots \sum_{b_n \in \{0,1\}} P_{\phi}(b_1, \dots b_n)$$

$$\# \phi(X_1, \dots, X_{i-1}) = \Sigma_{b_i \in \{0,1\}} \dots \Sigma_{b_n \in \{0,1\}} P_{\phi}(X_1, \dots, X_{i-1}, b_i, \dots, b_n)$$

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Once we do that, we can plug in any element in F_{p} into our polynomial

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We ask the prover to provide this prime at the start of the protocol and the verifier can verify primalty in polynomial time.

A generic protocol to verify equations of the form

$$K = \sum_{b_{1} \in \{0,1\}} \dots \sum_{b_{n} \in \{0,1\}} g(X_{1}, \dots X_{n}) \qquad \dots eq(1)$$

Where g is any polynomial of small size and which can be evaluated in polynomial time.

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Obs: P_{ϕ} is a degree 3m polynomial it's size is of the order of the size of ϕ . It can also be easily evaluated in the same way we evaluate formulas on assignments. So we can use the sumcheck protocol.

Obs: $h(X_1) = \sum_{b_2 \in \{0,1\}} \dots \sum_{b_n \in \{0,1\}} g(X_1, b_2, \dots b_n)$

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Obs: g can be evaluated in polynomial time, however h cannot even be computed in polynomial time

Input: $g(X_1,...X_n)$, K

V: if n = 1, verify K = g(0) + g(1)

V: It asks the prover to send a polynomial h, as defined previously, a polynomial in X_1

P: sends a polynomial s

V: verify that s(0) + s(1) = K. Selects a random element from F_p , say a. It calculates s(a).

Recursively solve with the input as

 $g(a, X_2, \dots X_n)$ and s(a).

 $g(X_1,..,X_n)$

g(X₁,..X_n) -
g(X₁,..X_n) \rightarrow S₁(X₁)

g(X₁,..X_n) \implies $s_1(X_1)$





 $g(a_1, X_2, ..., X_n)$













. . .



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 $S_n(X_n)$





 $s_n(0)+s_n(1) == s_{n-1}(a_{n-1})$



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• How lucky does the prover need to be for the verifier to accept an incorrect string?

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Therefore the error probability is less than $3n^2/2^n$ which is less than $\frac{1}{3}$ for n>9

What's in IP?



TQBF ⊆ IP?

Definition: TQBF

TQBF = { $\Psi = Q_1 x_1 \dots Q_n x_n \phi(x_1, \dots, x_n) | \Psi = \text{True}, Q_i \text{ in } \{\exists, \forall\}, \text{ boolean formula } \phi$ }

$$\Psi = \forall x_1, \exists x_2, \forall x_3, \dots \exists x_n \ \phi(x_1, \dots, x_n) \in \mathsf{TQBF} \text{ iff}$$
$$\Pi_{b_1 \in \{0,1\}} \Sigma_{b_2 \in \{0,1\}} \Pi_{b_3 \in \{0,1\}} \dots \Sigma_{b_n \in \{0,1\}} \mathsf{P}_{\phi}(b_1, \dots, b_n) = 1$$

Where P_{ϕ} is the polynomial as defined before over F_2

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When we have a univariate polynomial over a variable quantified by \forall , we must check multiplicity, i.e, $s(0) \cdot s(1) = K$

• Unlike adding polynomials, multiplying polynomials increase the degree

• If we define $h(X_1)$ as defined previously:

$$h(X_1) = \sum_{b_2 \in \{0,1\}} \prod_{b_3 = \{0,1\}} \dots \sum_{b_n \in \{0,1\}} P_{\phi}(X_1, \dots b_n)$$

This can have degree at most 2ⁿ. Which cannot be sent from the prover to the verifier.

Obs:

$x^{k} = x$ in F_{2} for any k > 0

Linearization

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Any polynomial $p(X_1,...,X_n)$ can be converted to a *multilinear* polynomial $q(X_1,...,X_n)$ where

1. The degree of any variable in any term of q is at most 1

2.
$$p(a_1,...,a_n) = q(a_1,...,a_n)$$
 for any $a_1...,a_n \in \{0,1\}$

Definition: Linearization operator L

$$L_{i}(p) = X_{i} \cdot p(X_{1},...,X_{i-1},1,X_{i+1},...,X_{n}) + (1-X_{i}) \cdot p(X_{1},...,X_{i-1},0,X_{i+1},...,X_{n})$$

Defines a new polynomial such that

- 1. Degree of X_i in $L_i(p)$ is at most 1
- 2. $L_i(p)$ gives the same values as p for all binary inputs

Obs:
$$q = L_1(L_2(...L_n(p)...)))$$

Definition: ∀ operator for polynomials

$$\forall_{i} p(X_{1},...,X_{n}) = p(X_{1},...,X_{i-1},0,X_{i+1},...,X_{n}) \cdot p(X_{1},...,X_{i-1},1,X_{i+1},...,X_{n})$$

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The size of this expression is increased due to the addition of the linearization operator. The size will then be $O(n+1+2+...+n+|P_{\phi}|)$, which is still poly-size

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TQBF: g would be P_{ϕ} , t would be $o(n^3)$, and C would be 1

V: provide a polynomial equal to $R_2...R_tg(X_1,...X_n)$

P: returns a polynomial $s(X_1)$

V: 1) If $R_1 = \exists_1$ verify that s(0) + s(1) = C

2) If
$$R_1 = \forall_1$$
 verify that $s(0) \cdot s(1) = C$

3) If
$$R_1 = L_1$$
 and verify that $a \cdot s(1) + (1-a) \cdot s(0) = s(a)$

If all checks pass, pick a random element a, recursively prove that the polynomial $R_2...R_tg(a,...X_n) = s(a)$

 $\forall_1 L_1 \exists_2 L_1 L_2 \forall_3 \dots \exists_n L_1 L_2 \dots L_n g(X_1, \dots, X_n)$



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Modified Sumcheck protocol $\forall_1 L_1 \exists_2 L_1 L_2 \forall_3 \dots \exists_n L_1 L_2 \dots L_n g(X_1, \dots, X_n)$ $s_1(0) \cdot s_1(1) == C$ $L_1 \exists_2 L_1 L_2 \forall_3 \dots \exists_n L_1 L_2 \dots L_n g(a_1, X_2, \dots, X_n)$

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 $\exists_2 L_1 L_2 \forall_3 \dots \exists_n L_1 L_2 \dots L_n g(a_1, a_2, \dots X_n)$













 $S_t(X_n)$








Where's IP?



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• We don't need to restrict ourselves to one prover. If we could interact with multiple provers, we would get the class **MIP[BGK '88]**

• Note: Provers cannot talk to each other, they communicate only to the verifier on the transcript which everyone can see.

- What power does each prover give? More Provers => More Power?
 No.
- Theorem[BFL '91]: MIP = MIP[2] = NEXPTIME

• Replacing the BPP verifier with a BQP verifier in IP gives QIP[Wat '99]

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• We would be able to solve undecidable problems like the halting problem

IP = PSPACE Timeline

1985: AM, MA defined by Babai

1986: Goldwasser and Sipser show public coin private coin equivalence

1988: AM=AM[2] by BM, MIP is defined by BGKW

1989: IP is defined by GMR

1991: ZKP(NONISO in IP) by GMW, MIP=NEXP by BFL

1992: #3SAT in IP by LFKN, IP=PSPACE by Shamir, Simpler proof by Shen

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TL; DR

• Randomness+Interaction is the key, alone they are "weak"

• Supreme power is useless unless succinct

• Mapping to polynomials is a very powerful technique

