

# Complexity Classes of Counting Problems

Andreas-Nikolas Göbel

National Technical University of Athens, Greece

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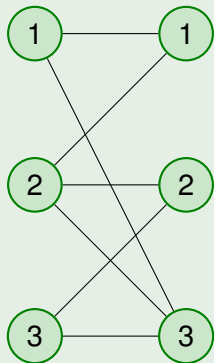
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  - Superclasses of #P
- 2 The Computational Power of #P
  - The Valiant Vazirani Theorem
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# Counting problems

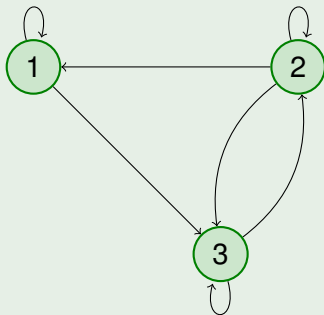
- **#SAT**: the function that counts the number of satisfying assignments of a given formula.
- **#HAMILTONPATHS**: the function that counts the number of Hamilton paths of a given graph.
- **#DIV**: the function that counts the number of divisors of a given number
- **#PRDIV**: the function that counts the number of **prime** divisors of a given number
- **#PERFECTMATCHINGS**: the function that counts the perfect matchings of a given bipartite graph.
- **#CYCLECOVER**: the function that counts the cycle covers of a given directed graph with self loops.
- **(0,1)-PERMANENT**: The permanent of a given matrix with elements from  $\{0, 1\}$

Remark: The last 3 are equivalent

## Perfect Matchings



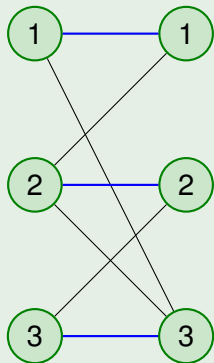
## Cycle Covers



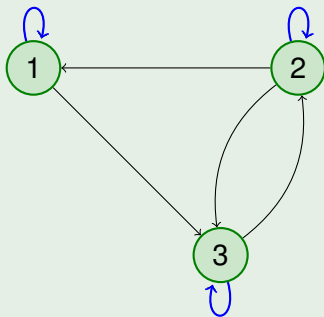
## Adjacency Matrix

1	0	1
1	1	1
0	1	1

## Perfect Matchings


 $(1,1), (2,2), (3,3)$ 

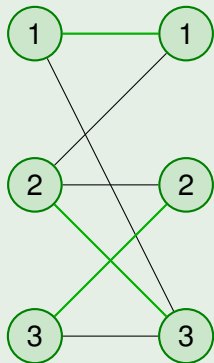
## Cycle Covers


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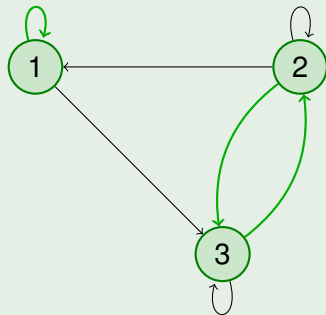
## Adjacency Matrix

1	0	1
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## Perfect Matchings


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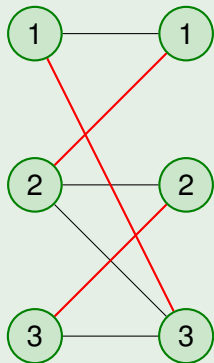
## Cycle Covers


 $(1,1), (2,3), (3,2)$ 

## Adjacency Matrix

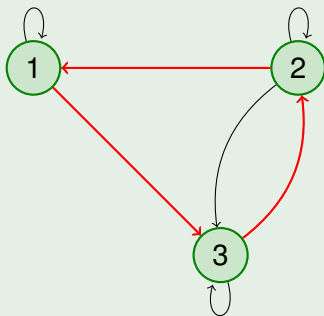
1	0	1
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## Perfect Matchings



(1,3),(2,1),(3,2)

## Cycle Covers



(1,3),(2,1),(3,2)

## Adjacency Matrix

1	0	1
1	1	1
0	1	1

## Definition (Valiant '79)

#P is the class of functions that can be computed by NDTM's of polynomial time complexity.

$$\#P = \{f : f(x) = \text{acc}_M(x)\}$$

Or

The class of functions that count the number of *witnesses* (or certificates) of NP problems.

**Def.**  $f \in \#P : \exists \text{ pred. } Q \in P, \forall x : f(x) = \#\{y \mid Q(x, y)\}$



# PP

## Definition (Wagner '86)

PP is the class of all languages  $L$  such that there exists a poly-NDTM  $M$  and an FP function  $f$  such that

$$L = \{x : \text{acc}_M(x) > f(x)\}$$

- Originally defined by Gill '74 as a probabilistic class
- From the above definition we can show:  $P^{\#P} = P^{PP}$
- $P^{\text{NP}[log]} \subseteq PP$  (Beigel, Hemaspaandra, and Wechsung '89)

### Definition (Wagner '89)

$C_{=P}$  is the class of all languages  $L$  such that there exists a poly-NDTM  $M$  and an FP function  $f$  such that

$$L = \{x : \text{acc}_M(x) = f(x)\}$$

### Definition (Papadimitriou and Zachos '82)

$\oplus P$  is the class of all languages  $L$  such that there exists a poly-NDTM  $M$  such that

$$L = \{x : \text{acc}_M(x) \text{ is odd}\}$$

### Definition (Allender '86)

For any language  $L$ ,  $L \in \text{FewP}$  if and only if there exist a NDTM  $M$  and a polynomial  $p$  such that:

$$\text{acc}_M(x) \leq p(|x|)$$

$$x \in L \Leftrightarrow \text{acc}_M(x) > 0$$

### Definition (Cai and Hemaspaandra '90)

For any language  $L$ ,  $L \in \text{Few}$  if and only if there exist a NDTM  $M$  and a polynomial  $p$  and a polynomial time computable predicate  $A(x, y)$  such that:

$$\text{acc}_M(x) \leq p(|x|)$$

$$x \in L \Leftrightarrow A(x, \text{acc}_M(x))$$

- $\oplus P^{\oplus P} = \oplus P$  (PZ '82)
- $\text{Mod}_k P^{\text{Mod}_k P} = \text{Mod}_k P$  holds if  $k$  is a prime (BGH '90)
- $\text{FewP} \subseteq \text{NP}$
- This is not known for Few but  $\text{Few} \subseteq P^{\text{NP}[\log]}$
- $\text{Few} \subseteq \oplus P$  (CH '90),  
 $\text{Few} \subseteq C=P$  (KSTT '89),  
 $\text{Few} \subseteq \text{Mod}_k P$ , for each prime  $k$  (BGH '90)
- Few is low for the classes PP, C=P and  $\oplus P$  (KSTT '89)
  - That is  $\text{PP}^{\text{Few}} = \text{PP}$

# Only Positive?

- The counting classes defined so far may only contain positive functions.
- For a NDTM  $M$  we denote with  $\overline{M}$  the machine identical to  $M$  but with the accepting and rejecting states interchanged
- $\text{gap}_M(x) = \text{acc}_M(x) - \text{acc}_{\overline{M}}(x)$

## Definition

$\text{GapP} = \{\text{gap}_M : M \text{ is a poly-NDTM}\}$

- $\#P \subseteq \text{GapP} = \#P - \#P = \#P - \text{FP} = \text{FP} - \#P$

# SPP

## Definition

SPP is the class of all languages  $L$ , such that there exists a poly-NDTM  $M$ , such that for all  $x$

$$x \in L \Rightarrow \text{gap}_M(x) = 1,$$

$$x \notin L \Rightarrow \text{gap}_M(x) = 0.$$

Köbler, Schöning and Torán ('92) showed that Graph Automorphism is in SPP

Arvind and Kurur ('02) showed that Graph Isomorphism is in SPP

# Valiant's Framework

## Definition

For every complexity class  $\mathcal{C}$  of decision problems we define  $\#\mathcal{C} = \bigcup_{A \in \mathcal{C}} (\#P)^A$ , where  $\#P^A$  is the collection of all functions that count the accepting paths of polynomially bounded NDTM's having  $A$  as their oracle.

$\#\mathcal{C} = \#\text{co}\mathcal{C}$  holds for every complexity class

# Span and Operators

## Definition (Hemaspaandra Vollmer '95)

For any class  $\mathcal{C}$ , define  $\# \cdot \mathcal{C}$  to be the class of functions  $f$ , such that for some  $\mathcal{C}$ -computable 2-ary predicate (relation)  $R$  and some polynomial  $p$ , for every string  $x$  it holds that:

$$f(x) = \|\{y : p(|x|) = |y| \text{ and } R(x, y)\}\|,$$

where “ $\|A\|$ ” denotes the cardinality of the set  $A$ .

## Definition (Köbler, Shöning, Torán '89)

For a non-deterministic transducer  $M$  define the function  $\text{span}_M : \Sigma^* \rightarrow \mathbb{N}$  such that  $\text{span}_M(x)$  is the number of different valid outputs that occur in the nondeterministic computation tree induced by  $M$  on input  $x$ .

Define  $\text{SpanP} = \{f : f = \text{span}_M \text{ and } M \text{ is a poly-NDTM}\}$



## Definition

$\text{span}_{M-N}(x)$  is the number of different outputs that  $M(x)$  can produce and  $N(x)$  can't.

## Lemma

$\#NP = \{f : f = \text{span}_{M-N}\}, M, N \text{ PNTM's}$

$$\#NP \subseteq \# \cdot NP - \# \cdot NP$$

# Some Results

- $\#NP = \# \cdot P^{NP}$  (from definitions)
- $\#NP = (\#P)^{NP[1]}$  (KST '89)
- $\#P \subseteq \# \cdot NP \subseteq \#NP$  (KST '89)
- $\#P = \# \cdot NP$  if and only if  $UP = NP$  (KST '89)
- $\# \cdot NP = \#NP$  if and only if  $NP = coNP$  (KST '89)

# Toda's Result (Ph.D.'92)

## Theorem

$$\# \cdot \text{coNP} = \# \text{NP}$$

- Generalizing to the #PH we have:

$$\# \cdot \Sigma_k^p \subseteq \# \Sigma_k^p = \# \cdot \Pi_k^p$$

- Thus,  $\# \text{PH} = \bigcup \# \Sigma_k^p = \bigcup \# \cdot \Pi_k^p = \# \cdot \text{PH}$

# Valian-Vazirani Theorem

- For every known NP-complete problem the number of solutions of its instances varies from zero to exponentially many.
- Does this cause the inherent intractability of these problems?

## Theorem (Valiant Vazirani '86)

*There exists a probabilistic polynomial time algorithm  $f$  s.t. for every  $n$ -variable Boolean formula  $\varphi$*

$$\varphi \in \text{SAT} \Rightarrow \Pr[f(\varphi) \in \text{USAT}] \geq \frac{1}{8n}$$

$$\varphi \notin \text{SAT} \Rightarrow \Pr[f(\varphi) \in \text{SAT}] = 0$$

Therefore the above answer is no unless  $NP \stackrel{\square}{=} RP$

# Pairwise Independent Hash Functions

## Definition (Pairwise Independent Hash Functions)

Let  $\mathcal{H}_{n,k}$  be a collection of functions from  $\{0, 1\}^n$  to  $\{0, 1\}^k$ . We say that  $\mathcal{H}_{n,k}$  is pairwise independent if for every  $x, x' \in \{0, 1\}^n$  with  $x \neq x'$  and for every  $y \neq y' \in \{0, 1\}^k$ ,

$$\Pr_{h \in_R \mathcal{H}_{n,k}} [h(x) = y \wedge h(x') = y'] = 2^{-2k}$$

Lemma: Let  $\mathcal{H}_{n,k}$  be a pairwise independent hash function collection from  $\{0, 1\}^n$  to  $\{0, 1\}^k$  and  $S \subseteq \{0, 1\}^n$  s.t.  $2^{k-2} \leq |S| \leq 2^{k-1}$ . Then

$$\Pr_{h \in_R \mathcal{H}_{n,k}} [\text{there is a unique } x \in S \text{ satisfying } h(x) = 0^k] \geq \frac{1}{8}$$

proof on board

# Proof of Valiant Vazirani Theorem

- Given  $\varphi$  on  $n$  variables, choose  $k$  at random from  $\{2, \dots, n+1\}$  and a random hash function  $h \in_R \mathcal{H}_{n,k}$ .
- The statement  $\exists_{x \in \{0,1\}^n} \varphi(x) \wedge (h(x) = 0^k)$  is false if  $\varphi$  is unsatisfiable, and with probability  $1/8n$  has a unique satisfying assignment if  $\varphi$  is satisfiable.
  - If  $S$  is the set of satisfying assignments of  $\varphi$ , with probability  $1/n$ ,  $k$  satisfies  $2^{k-2} \leq |S| \leq 2^{k-1}$ .
  - With probability  $1/8$  there is a unique  $x$  such that  $\varphi(x) \wedge h(x) = 0^k$
- The implementation is based on Cook's reduction, and expresses the deterministic computation inside the  $\exists$  sign as a formula  $\tau$  on variables  $x, y \in \{0, 1\}^{\text{poly}(n)}$ , s.t.  $h(x) = 0$  iff there exists a unique  $y$  such that  $\tau(x, y) = 1$ . Output

$$\psi = \varphi(x) \wedge \tau(x, y)$$

# What is the complexity of #P compared to Decision Classes?

Theorem (Papadimitriou Zachos '82)

$$P^{NP[\log]} \subseteq P^{\#P[1]}$$

Theorem (Toda '91)

$$PH \subseteq P^{\#P}$$

# Proof

## Lemma (Randomized Reduction from PH to $\oplus$ SAT)

*There exists a prob. poly-time algo  $A$  s.t. given a parameter  $m$  and any QBF  $\psi$  of size  $n$  with  $c$  levels of alternations, runs in  $\text{poly}(n, m)$  and satisfies:*

$$\psi \text{ is true} \Rightarrow \Pr[A(\psi) \in \oplus\text{SAT}] \geq 1 - 2^{-m}$$

$$\psi \text{ is false} \Rightarrow \Pr[A(\psi) \in \oplus\text{SAT}] \leq 2^{-m}$$

### Proof on board

- By derandomizing the above lemma the proof of Toda's theorem is concluded.



# Hard To Count - Easy to Decide

- Surprisingly enough, some problems in P have counting versions that are complete for #P under less restrictive reductions.
  - Cook reductions** ( $f \leq_T^p g : f \in \text{FP}^g$  — aka poly-time Turing).
- Examples: #PERFECT MATCHINGS, #DNFSAT, #MONSAT
- They cannot be complete for #P under parsimonious reductions unless  $P = \text{NP}$ .
- Note that #P is not closed (under likely assumptions) under Cook reductions.  
Therefore these problems *are Cook-complete also for superclasses (and subclasses!) of #P.*
- Hard to count - easy to decide problems neither are well represented by #P, nor are well classified by means of Cook reductions.

# Further Motivation for Studying HCED Problems

- Three degrees of approximability within problems of #P [DGGJ'00]:
  - Solvable by an *FPRAS*:  
#PERFECT MATCHINGS, #DNFSAT, ...
  - AP-interreducible with SAT:  
SAT, #IS, ...
  - An Intermediate Class (AP-Interreducible with #BIS)  
#BIS, ...

# Hard to Count Easy to Decide

- #P contains counting versions of known NP problems
- there exist other #P problems with decision version in P

## Definition (Pagourtzis '01)

Let #PE be the class that contains functions of #P whose related language is in P.

## Definition (Kiayias, Pagourtzis, Sharma and Zachos)

$\text{TotP} = \{\text{tot}_M : M \text{ is a poly-NDTM}\}$

$\text{tot}_M(x) = (\text{The number of paths of } M \text{ on input } x) - 1$

# Inclusions among #P, #PE, and TotP

- $\#P \subseteq \text{TotP} - \text{FP}$
- $\text{FP} \subseteq \text{TotP} \subseteq \#PE \subseteq \#P$ .  
These inclusions are proper unless  $P = NP$
- From these propositions we get the following Corollaries:
  - TotP, #PE and #P are not Karp equivalent unless  $P=NP$ .
  - The above classes are Cook[1] irreducible:

$$\text{FP}^{\text{TotP}[1]} = \text{FP}^{\#PE[1]} = \text{FP}^{\#P[1]}$$

- Combining the latter with Toda's result we have

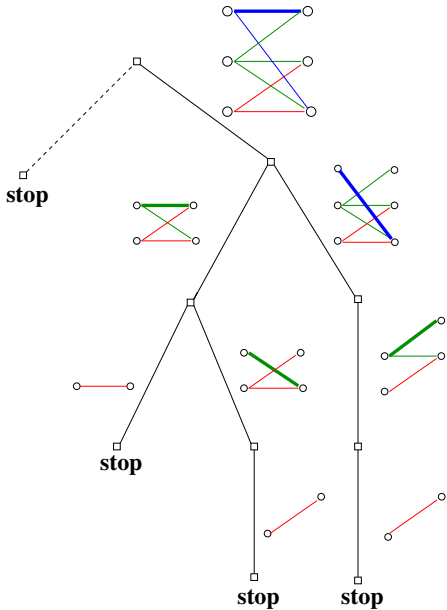
$$\text{PH} \subseteq \text{P}^{\text{TotP}[1]} = \text{P}^{\#PE[1]}$$

# #PERFECT MATCHINGS is in TotP

*Sketch of the proof:*

- It can be decided in polynomial time whether a graph  $G$  has any perfect matchings.
- Let  $e = (v, u)$  be an edge of  $G$ . Then, the set of perfect matchings of  $G$  can be **partitioned into two subsets**,  $S_0$  and  $S_1$ , where  $S_0$  consists of those perfect matchings that match  $u$  and  $v$  through  $e$ , and  $S_1$  consists of the remaining perfect matchings.
- Each  $S_i$  has the same cardinality with the set of perfect matchings of an appropriate subgraph  $G_i$  of  $G$ .

Path Counting



# Self-reducibility of #DNF-SAT

Select a variable  $x$  and construct formulae

$$\phi_0 := \phi|_{x=0},$$

$$\phi_1 := \phi|_{x=1}. \text{ Clearly:}$$

$$[\#DNFSAT(\phi) = \#DNFSAT(\phi_0) + \#DNFSAT(\phi_1)]$$

## Corollary

$\#DNFSAT$  is in TotP.

# Self-reducibility for well-known problems that are therefore in TotP

- #PERFECTMATCHINGS (equiv. PERMANENT and CYCLECOVER)
- #DNF-SAT
- #2-SAT
- #NONCLIQUES
- #NONINDSETS
- #INDSETSALL
- RANKING

*They are all Cook[1]-complete for TotP.*



# TotP = Karp-closure( $\#PE_{SR}$ )

## Self Reducibility

A function  $f : \Sigma^* \rightarrow \mathbb{N}$  is called **poly-time self-reducible**, if there exist polynomials  $r$  and  $q$  and polynomial time computable functions  $h : \Sigma^* \rightarrow \Sigma^*$ ,  $g : \Sigma^* \rightarrow \mathbb{N}$  and  $t : \Sigma^* \rightarrow \mathbb{N}$ , such that for all  $x \in \Sigma^*$

- 1  $f(x) = t(x) + \sum_{i=0}^{r(|x|)} g(x, i)f(h(x, i))$ , that is,  $f$  can be processed recursively by reducing  $x$  to  $h(x, i)$ , ( $0 \leq i \leq r(|x|)$ ), and
- 2 the recursion terminates after at most polynomial depth (that is, the value of  $f$  on instance  $h(\dots h(h(x, i_1), i_2) \dots, i_{q(|x|)})$  can be computed deterministically in polynomial time).

## Theorem (Pagourtzis, Zachos)

**TotP** is exactly the closure under Karp reductions of  $\#PE_{SR}$

# Some Definitions

- $A$  is a Partial Order ( $x \leq_A y$ )
  - reflexive
  - antisymmetric
  - transitive
- $A$  is a Total Order
  - partial order
  - all strings comparable
- $A$  is a p-order
 

$\exists q, \forall x, y$  with  $x <_A y$  :  $|x| \leq q(|y|)$   
 (i.e. lengths of all strings are polynomially related)
- $A$  has Efficient Adjacency Checks (a feasibility constraint)  
 We can check efficiently (in P) whether:  
 $(x <_A y) \wedge (\nexists z : x <_A z <_A y)$

# Interval-size Function Classes

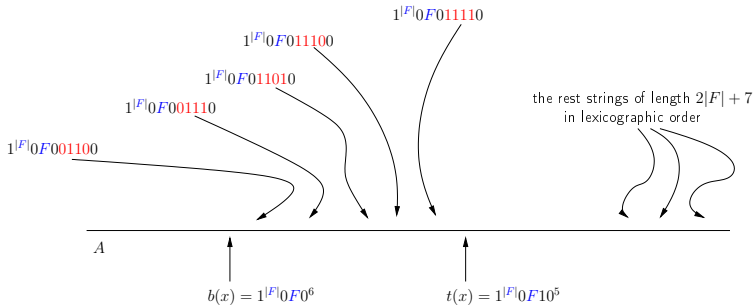
## Definition

- $IF_p$  ( $IF_t$ ) is the class of all functions  $f : \Sigma^* \rightarrow \mathbb{N}$  for which there exists a partial (total)  $p$ -order  $A \in \mathbf{P}$ , with  $f(x) = \|\{z : l(x) <_A z <_A u(x)\}\|$ , for every  $x \in \Sigma^*$ , where  $l, u \in FP$ .
- $IF_p^{\prec}$  ( $IF_t^{\prec}$ ) is the class of all functions  $f : \Sigma^* \rightarrow \mathbb{N}$  for which there exists a partial (total)  $p$ -order  $A \in \mathbf{P}$ , with  $f(x) = \|\{z : l(x) <_A z <_A u(x)\}\|$ , for every  $x \in \Sigma^*$ , where  $l, u \in FP$ , and  $A$  has **efficient adjacency checks**.
- $IF_p^{\prec}$  contains **#DIV** and **#PRDIV**
- $IF_t^{\prec}$  contains **#MONSAT** (also **Cook-complete for this class**).  
(A monotone CNF formula contains no  $\neg$ )

Interval Size Functions

# Example: an $IF_t^<$ computation for #MONSAT

$$F = \text{enc}[(x_0 \vee x_2) \wedge (x_0 \vee x_1 \vee x_3) \wedge (x_1) \wedge (x_2 \vee x_3)]$$



# Interval-size Function Classes (cont.)

## Theorem (Hemaspaandra, Homan, Kosub, Wagner '01)

For any function  $f$  the following are equivalent:

- 1  $f \in \#P$ .
- 2 There exist a partial  $p$ -order  $A \in \mathbf{P}$  with  $f(x) = \|\{z : l(x) <_A z <_A u(x)\}\|$ , for all  $x \in \Sigma^*$ , for some  $l, u \in \text{FP}$ .
- 3 There exist a total  $p$ -order  $A \in \mathbf{P}$  with  $f(x) = \|\{z : l(x) <_A z <_A u(x)\}\|$ , for all  $x \in \Sigma^*$ , for some  $l, u \in \text{FP}$  with  $l(x) <_A u(x)$ .

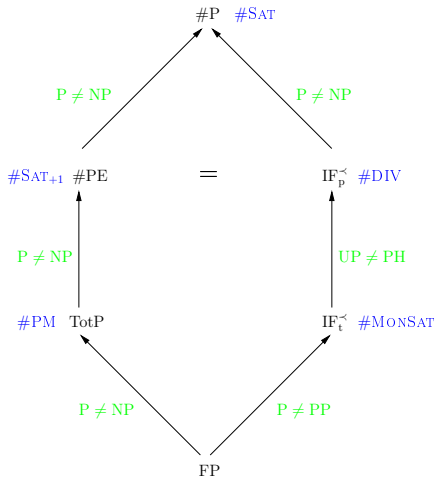
Corollary:  $\#P = \text{IF}_t = \text{IF}_p$ .

# Some Inclusions

- $\text{IF}_p^{\succ} = \#PE$  [Hemaspaandra et. al.'01]
- $\#P \subseteq \text{IF}_t^{\succ} - \text{FP}$
- $\text{FP}^{\text{IF}_t^{\succ}} = \text{FP}^{\text{IF}_p^{\succ}} = \text{FP}\#P$
- $\text{FP} \subseteq \text{IF}_t^{\succ} \subseteq \text{IF}_p^{\succ} \subseteq \#P = \text{IF}_t = \text{IF}_p$
- The inclusion  $\text{FP} \subseteq \text{IF}_t^{\succ}$  is proper unless  $\text{FP} = \#P$ .

Connecting the two Approaches

# Summarizing Earlier Results



## Other Feasibility Constraints

Consider other polynomial time feasibility constraints, besides efficient adjacency checks.

**Def:**  $IF_t^{LN}$ : **Lexicographical Nearest Function** (Given a string compute the lex-nearest string within a defined interval)

**Theorem:**  $IF_t^{LN} = \text{TotP}$

**Def:**  $IF_t^{\text{med}}$ : **median function**.

(**Fact:**  $IF_t^{\text{med}} = \text{FP}$ )

**Def:**  $IF_t^{\text{rmed}}$ : **“relaxed” median function** (there is a whole family of such).

(**Fact:** Contains the problem  $\#SAT_{+2^n}$ )

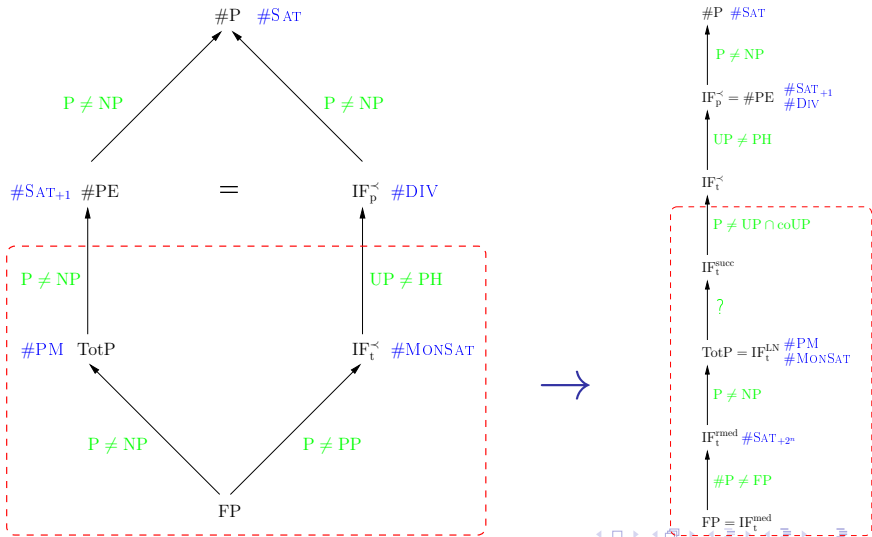
**Corollary:**

$$\text{FP} = IF_t^{\text{med}} \subseteq IF_t^{\text{rmed}} \subseteq IF_t^{LN} = \text{TotP}$$



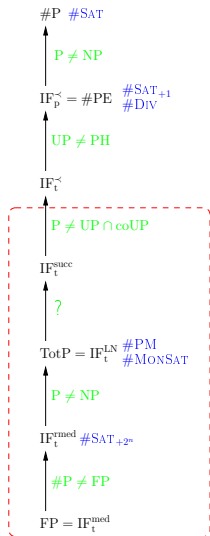
Connecting the two Approaches

# The Results [Bampas G. Pagourtzis Tentis '09]



Connecting the two Approaches

# The Results (cont.)



- Interval size characterizations:  
 $TotP = IF_t^{LN}$  (also  $FP = IF_t^{med}$ ).
- $TotP \subseteq IF_t^{<}$ ; the inclusion is proper unless  $P = UP \cap coUP$ .
- A new interval size class,  $IF_t^{rmed}$ , s.t.  
 $FP \subseteq IF_t^{rmed} \subseteq TotP$ , the inclusions being proper unless  $FP = \#P$  and  $P = NP$  resp.
- $\#SAT_{+2^n} \in IF_t^{rmed}$ . Does  $IF_t^{rmed}$  contain natural “easy to decide - hard to count” problems?

# Interval size characterizations

- The main idea is to consider new polynomial time feasibility constraints, other than efficient adjacency checks.
- For FP we consider the median function  $\text{med}_A(x, y)$
- For TotP:
  - Lexicographical Nearest Function on a p-order  $A$ :  
 $\text{LN}_A(x, y, w)$  returns the string  $z \in (x, y)_A$ , which is the lex nearest to  $w$  among strings in  $(x, y)_A$ .
  - Captures the property that given a NPTM  $M$  and a string of non-deterministic choices, we can find the lex nearest string that encodes a computation path of  $M$ .
  - All #PE self reducible problems (i.e. TotP) can be expressed as interval size functions with lex-nearest function in P
  - *And vice versa*

$$\text{TotP} = \text{IF}_t^{\text{LN}}$$

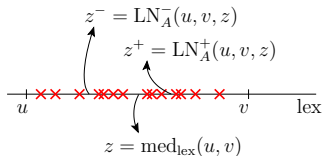
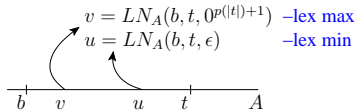
- $\text{TotP} \subseteq \text{IF}_t^{\text{LN}}$ 
  - Let  $M$  be a TotP machine.
  - The boundary functions  $b, t$  are set according to  $p(|x|)$ , where  $p$  is the polynomial bounding the length of  $M$ 's computation.
  - An appropriate order  $A$  is defined, such that for each path of  $M$ , a string is contained in  $(b(x), t(x))_A$  iff it encodes a valid computation path of  $M(x)$ .
  - The computation of  $\text{LN}_A$  is based on the property that given a string  $x$ , we can verify efficiently whether it encodes a computation path of  $M$  and if not, we can efficiently find the encoding of a path  $y$ , that is lex-nearest to  $x$ .

# TotP = IF<sub>t</sub><sup>LN</sup> (concl.)

- IF<sub>t</sub><sup>LN</sup> ⊆ TotP
  - We construct an NPTM  $M$  s.t. on input  $x$ ,  $\text{tot}_M(x) = \|(b(x), t(x))_A\|$ .
  - We use the LN<sub>A</sub> function in order to re-arrange strings in  $(b(x), t(x))_A$  lexicographically.
  - The lex-first and last strings of  $(b(x), t(x))_A$ , say  $f, l$  are computed first.
  - Then, the lex-nearest to the lex-median of  $(f, l)_A$  is computed, say  $m$ .
  - The process is repeated recursively by splitting (into two computation paths) as long as  $(f, m)$  and  $(m, l)$  are nonempty.

$\text{IF}_t^{\text{LN}} \subseteq \text{TotP}$ 

- If we can compute  $\text{LN}_A$  efficiently then we can also compute  $\text{LN}_A^+$  and  $\text{LN}_A^-$  efficiently.
- We use  $\text{LN}_A$  function in order to re-arrange strings in  $(b(x), t(x))_A$  lexicographically.



# Functions-to-Languages Operators

- We define class operators that produce appropriate decision versions of counting problems.
- We observe that by using efficient adjacency checks ( $\text{IF}_t^{\prec}$ ) we can only tell (efficiently) whether there are any strings in the interval.
- On the other hand, in a TotP computation we can decide whether there are more than one paths.

## Definition

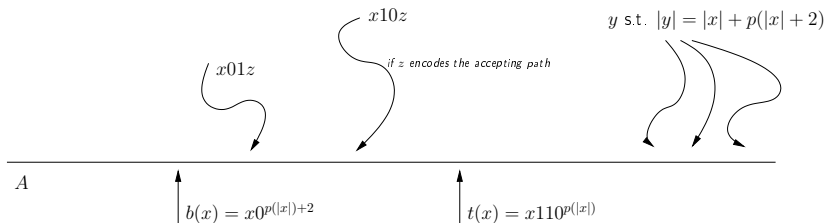
Let  $\mathcal{F}$  be a function class; then  $\mathcal{C}_{>1} \cdot \mathcal{F}$  is the following class of languages:

$$\mathcal{C}_{>1} \cdot \mathcal{F} = \{L \mid \exists f \in \mathcal{F} \forall x (x \in L \iff f(x) > 1)\} .$$

## Connecting the two Approaches

TotP  $\not\subseteq$  IF<sub>t</sub><sup>↖</sup> if P ≠ UP ∩ coUP

- $\mathcal{C}_{>1} \cdot \text{TotP} \subseteq \text{P}$
- $\text{UP} \cap \text{coUP} \subseteq \mathcal{C}_{>1} \cdot \text{IF}_t^{\swarrow}$



- If  $\text{IF}_t^{\swarrow} \subseteq \text{TotP}$ , then  $\text{UP} \cap \text{coUP} \subseteq \mathcal{C}_{>1} \cdot \text{IF}_t^{\swarrow} \subseteq \mathcal{C}_{>1} \cdot \text{TotP} \subseteq \text{P}$



# Open Problems

- Can the Karp closure of **#PERFECT MATCHINGS**, **#DNFSAT**, etc., be described in terms of interval size functions? In terms of path counting functions?
- Are there complete problems for the studied classes under Karp reductions? Under other suitable reductions?
- Can the approximation subclasses of #P (DGGJ'00) be related to interval size functions or to path counting functions?

THANK YOU!!