# Enumeration: logical and algebraic approach 

Yann Strozecki

Université Paris Sud - Paris 11
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Introduction to Enumeration

Enumeration and logic

Enumeration and polynomials

## Enumeration problems

Polynomially balanced predicate $A(x, y)$, decidable in polynomial time.

- $\exists$ ? $y A(x, y)$ : decision problem (class NP)
- $\sharp\{y \mid A(x, y)\}$ : counting problem (class $\sharp \mathrm{P}$ )
- $\{y \mid A(x, y)\}$ : enumeration problem (class EnumP)


## Perfect matching: <br> - The decision problem is to decide if there is a perfect matching. <br> - The counting problem is to count the number of perfect matchings.

-The enumeration problem is to list every perfect matching

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## Example

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## Time complexity measures for enumeration

1. the total time related to the number of solutions

- polynomial total time: TotalP

2. the delay

- incremental polynomial time: IncP (Circuits of a matroid)
- polynomial delay: DelayP (Perfect Matching [Uno])
- Constant or linear delay
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## Enumeration problems

$R$ : polynomially balanced binary predicate
EnUM $\cdot R$
$\begin{array}{ll}\text { Input: } & x \in \mathcal{I} \\ \text { Output: } & \text { an enumeration of elements in } R(x)=\{y \mid R(x, y)\}\end{array}$

## Definition

The problem Enum• $R$ belongs to the class Delay $(g, f)$ if there exists an enumeration algorithm that computes EnUm. $R$ such that, for all input $x$ :

- Preprocessing in time $O(g(|x|))$,
- Solutions $y \in R(x)$ are computed successively without repetition with a delay $O(f(|x|))$
$\operatorname{Constant}-\operatorname{Delay}=\bigcup_{k} \operatorname{Delay}\left(n^{k}, 1\right)$.


## Enumeration complexity classes

Separation:

QueryP $\subsetneq \mathbf{S D e l a y P} \subseteq$ Delay $\mathbf{P} \subseteq \operatorname{Inc} \mathbf{P} \subsetneq$ Total $\mathbf{P} \subsetneq$ EnumP.

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> If $\mathrm{P} \neq \mathrm{NP}$ then the classes DelayP, IncP and TotalP are not stable by subtraction.

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## Meta-algorithms for enumeration and CSP

> Proposition (Creignou, Hebrard'97)
> The problem Enum•SAT $(\mathcal{C})$ is in DelayP when $\mathcal{C}$ is one of the following classes: Horn formulas, anti-Horn formulas, affine formulas, bijunctive (2CNF) formulas.

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## Logic in half a slide

## First order logic(FO):

- Variables: $x, y, z \ldots$
- The language $\sigma$, relations and functions: $R(x, y), f(z)$
- Unary and binary connectors: $\wedge, \vee, \neg$
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## Theorem (Goldberg)

For almost all first order graph property $\varphi$, the graphs of size $n$ which satisfies $\varphi$ can be enumerated with polynomial delay in $n$.

## Enumeration problem defined by a formula

Second order logic(SO):
Second order variable: $\mathbf{T}$, denotes unknown relation over the domain.

Let $\Phi(\mathbf{z}, \mathbf{T})$ be a first order formula with free first and second
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```
Enum·\Phi
    Input: A }\sigma\mathrm{ -structure }\mathcal{S
    Output: }\quad\Phi(\mathcal{S})={(\mp@subsup{\mathbf{z}}{}{*},\mp@subsup{\mathbf{T}}{}{*}):(\mathcal{S},\mp@subsup{\mathbf{z}}{}{*},\mp@subsup{\mathbf{T}}{}{*})\models\Phi(\mathbf{z},\mathbf{T})
```

Let $\mathscr{F}$ be a subclass of first order formulas. We denote by Enum $\cdot \mathscr{F}$ the collection of problems Enum $\Phi$ for $\Phi \in \mathscr{F}$.

## Example

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Independent sets:

$$
I S(T) \equiv \forall x \forall y T(x) \wedge T(y) \Rightarrow \neg E(x, y)
$$

## Example

Hitting sets (vertex covers) of a hypergraph represented by the incidence structure $\langle D,\{V, E, R\}\rangle$.

$$
H S(T) \equiv \forall x(T(x) \Rightarrow V(x)) \wedge \forall y \exists x E(y) \Rightarrow(T(x) \wedge R(x, y))
$$

## First-order queries with free second order variables

## This presentation

- FO queries with free second-order variables
- Data complexity: the query is fixed
- The complexity in term of the size of the input structure's domain
- Quantifier depth as a parameter: EnUm• $\Sigma_{1}$


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- Quantifier depth as a parameter: EnUm. $\Sigma_{1}$
- Enum•IS $\in$ Enum• $\Pi_{1}$ and Enum•HS $\in$ Enum• $\Pi_{2}$


## Previous results

1. Only first-order free variables and bounded degree structures. Durand-Grandjean'07, Lindell'08, Kazana-Segoufin'10: linear preprocessing + constant delay.
2. Only first-order free variables and acyclic conjunctive formula. Bagan-Durand-Grandjean'07: linear preprocessing + linear

## Example

Enumeration of the $k$-cliques of a graph of bounded degree.

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Monadic second order formula and bounded tree-width structure Bagan, Courcelle 2009: almost linear preprocessing

## Example

Typical database query. Simple paths of length $k$.

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## Example

Enumeration of the cliques of a bounded tree-width graph.

## A hierarchy result for counting functions

From a formula $\Phi(\mathbf{z}, \mathbf{T})$, one defines the counting function:

$$
\# \Phi: \mathcal{S} \mapsto|\Phi(\mathcal{S})|
$$

## Theorem (Saluja, Subrahmanyam, Thakur 1995) <br> On linearly ordered structures: $\# \Sigma_{0} \subsetneq \# \Sigma_{1} \subsetneq \# \Pi_{1} \subsetneq \# \Sigma_{2} \subsetneq \# \Pi_{2}=\sharp P$.

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## Corollary

On linearly ordered structures:
Enum $\cdot \Sigma_{0} \subsetneq$ Enum $\cdot \Sigma_{1} \subsetneq$ Enum $\cdot \Pi_{1} \subsetneq$ Enum $\cdot \Sigma_{2} \subsetneq$ Enum $\cdot \Pi_{2}$.

## The first level: Enum• $\Sigma_{0}$

## Theorem

For $\varphi \in \Sigma_{0}$, Enum $\cdot \varphi$ can be enumerated with preprocessing $O\left(|D|^{k}\right)$ and delay $O(1)$ where $k$ is the number of free first order variables of $\varphi$ and $D$ is the domain of the input structure.

## Simple ingredients:

1. Transformation of a f.o. formula $\Phi(\mathbf{z}, T)$ into a propositional formula:

- Try all values for first order variables: $\Phi\left(\mathbf{z}^{*}, T\right)$.
- Replace the atomic formulas by their truth value.
- Obtain a propositional formula with variables $T(\mathbf{w})$.

2. Gray Code Enumeration.

## Bounded degree structure

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Theorem
Let $d \in \mathbb{N}$, on $d$-degree bounded input structures,
$\operatorname{Enum} \cdot \Sigma_{0} \in \operatorname{DELAY}(|D|, 1)$ where $D$ is the domain of the input structure.

Idea of proof:

- Another transformation: $\Phi(\mathbf{z}, T)$ seen as a propositional formula whose variables are the atoms of $\Phi$.
- From each solution, create a quantifier free formula without free second order variables.
- Enumerate the solutions of this formula thanks to [DG 2007]


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## Second level: Enum $\cdot \Sigma_{1}$

## Theorem

Enum $\cdot \Sigma_{1} \subseteq$ DelayP. More precisely, EnUm $\cdot \Sigma_{1}$ can be computed with precomputation $O\left(|D|^{h+k}\right)$ and delay $O\left(|D|^{k}\right)$ where $h$ is the number of free first order variables of the formula, $k$ the number of existentially quantified variables and $D$ is the domain of the input structure.

Idea of Proof: $\Phi(\mathbf{y}, T)=\exists \mathbf{x} \varphi(\mathbf{x}, \mathbf{y}, T)$

- Substitute values for $\mathbf{x}$. Collection of formulas of the form:

$$
\varphi\left(\mathbf{x}^{*}, \mathbf{y}, T\right)
$$

- Need to enumerate the (non necessarily disjoint) union.


## The case Enum $\cdot \Pi_{1}$

## Proposition

Unless $\mathrm{P}=\mathrm{NP}$, there is no polynomial delay algorithm for Enum• $\Pi_{1}$.

Proof Direct encoding of SAT.

Hardness even:

- on the class of bounded degree structure
- if all clauses but one have at most two occurences of a second-order free variable


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Corollary
Let }\Phi(\mathbf{z},T)\mathrm{ be a formula, such that, for all }\sigma\mathrm{ structures, all propositional formulas \(\tilde{\Phi}\) are either Horn, anti-Horn, affine or bijunctive. Then Enum• \(\Phi \subseteq\) DELAYP.
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Example: independent sets and hitting sets.

## Conlusion and open problems

Enum $\cdot \Sigma_{0} \subsetneq$ Enum $\cdot \Sigma_{1} \subsetneq$ EnUm $\cdot \Pi_{1} \subsetneq$ EnUM $\cdot \Sigma_{2} \subsetneq$ Enum $\cdot \Pi_{2}=$ EnumP.

- Nice but small hierarchy.
- Other tractable classes above $\Sigma_{1}$ (optimization operator)?
- Efficient probabilistic enumeration procedure?


# Introduction to Enumeration 

## Enumeration and logic

Enumeration and polynomials

## Polynomial given by a black-box



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\begin{gathered}
X_{1}=1, X_{2}=2, X_{3}=1 \\
1 * 2+1 * 1+2+1 \\
\text { Output }=6
\end{gathered}
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$$
\begin{gathered}
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-1 * 1+-1 * 2+1+2 \\
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Easy to evaluate polynomials whose monomials represent interesting combinatorial objects.

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Let $P$ be a non zero polynomial with $n$ variables of total degree $D$, if $x_{1}, \ldots, x_{n}$ are randomly chosen in a set of integers $S$ of size $\frac{D}{\epsilon}$ then the probability that $P\left(x_{1}, \ldots, x_{n}\right)=0$ is bounded by $\epsilon$.

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Enumeration complexity: produce the monomials one at a time with a good delay.

## From finding a monomial to interpolation

Assume there is a procedure which returns a monomial of a polynomial $P$, then it can be used to interpolate $P$.

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- Yes for polynomial of fixed degree $d$. One can find the "highest" degree polynomial with $O\left(n^{2} D^{d-1}\right)$ calls.
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Open question: how to efficiently represent and compute the partial polynomial at each step? Easier with circuits, formulas, polynomials of low degree, over fixed finite fields ?

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Open question: how to efficiently represent and compute the partial polynomial at each step? Easier with circuits, formulas, polynomials of low degree, over fixed finite fields ?

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How to achieve a polynomial delay ?

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## Multilinear polynomials

Partial-Monomial
Input: a polynomial given as a black box and two sets of variables $L_{1}$ and $L_{2}$
Output: accept if there is a monomial in the polynomial in which no variables of $L_{1}$ appear, but all of those of $L_{2}$ do.

When the polynomial is multilinear, this problem can be solved by finding the degree of $P_{\bar{L}_{1}}$ with regard to $L_{2}$ : test if the degree is equal to $\left|L_{2}\right|$

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Let $P$ be a multilinear polynomial with $n$ variables and a total degree $D$. There is an algorithm which computes the set of monomials of $P$ with probability $1-\epsilon$ and a delay polynomial in $n, D$ and $\log (\epsilon)^{-1}$.

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## Comparison to other algorithms

|  | Ben-Or Tiwari | Zippel | KS | My Algorithm |
| :--- | :--- | :--- | :--- | :--- |
| Algorithm type | Deterministic | Probabilistic | Probabilistic | Probabilistic |
| Number of calls | $2 T$ | $t n D$ | $t n^{7} D^{4}$ | $t n D\left(n+\log \left(\epsilon^{-1}\right)\right)$ |
| Total time | Quadratic in $T$ | Quadratic in $t$ | Quadratic in $t$ | Linear in $t$ |
| Enumeration | Exponential | TotalPP | IncPP | DelayPP |
| Size of points | $T \log (n)$ | $\log \left(n T^{2} \epsilon^{-1}\right)$ | $\log \left(n D \epsilon^{-1}\right)$ | $\log (D)$ |

Figure: Comparison of interpolation algorithms on multilinear polynomials

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Strategy: relate the enumeration problem to some decision problem.

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## Proposition

The problem Partial-Monomial restricted to degree 2 polynomials is NP-hard.

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Thanks, thanks, thanks, thanks, thanks, thanks, thanks, thanks, thanks, thanks Let's all do enumeration

