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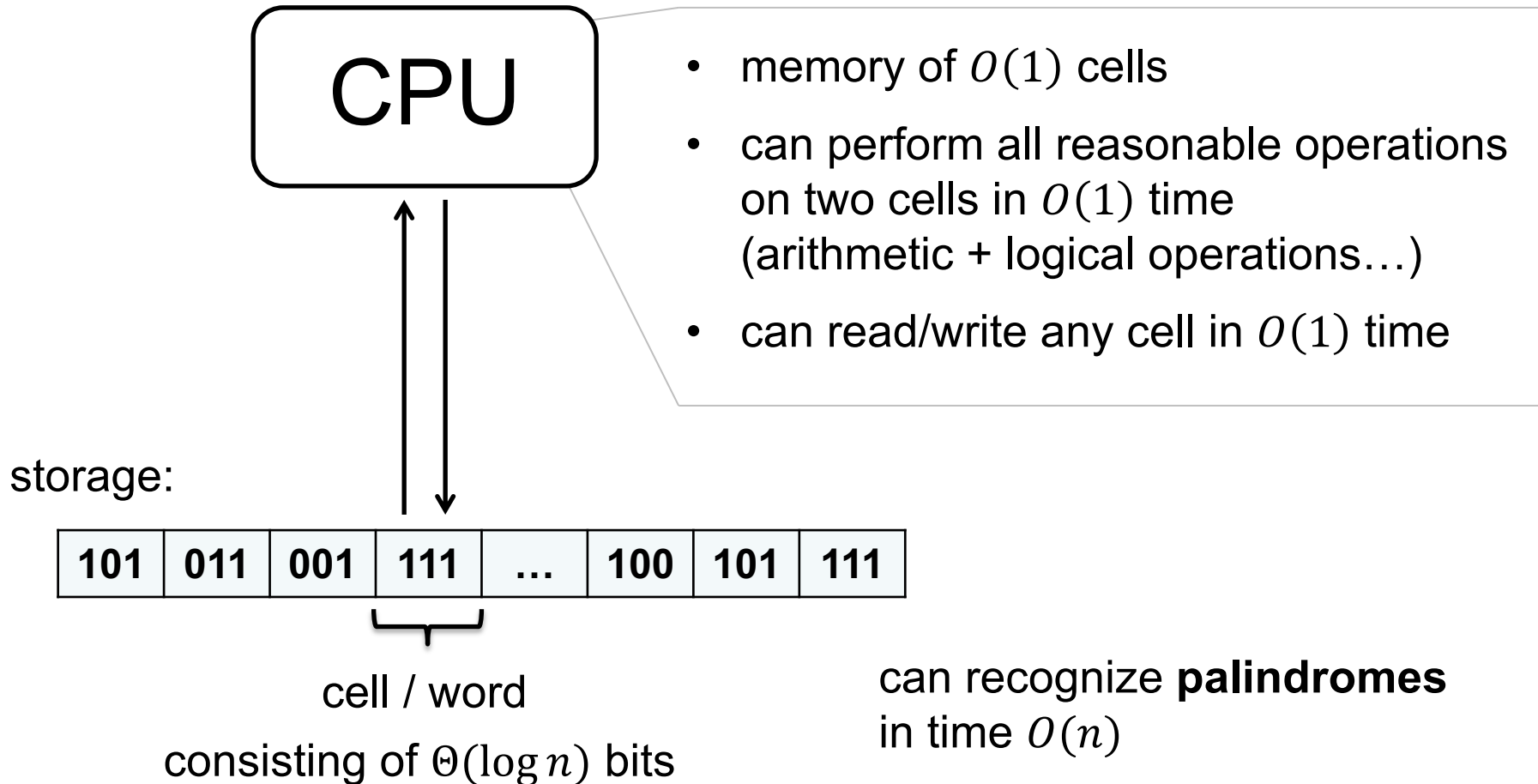
Fine-Grained Complexity - Hardness in P

Lecture 1: SETH and OV

Karl Bringmann

Machine Model

Random Access Machine (RAM):



can recognize **palindromes**
in time $O(n)$

the details do not matter!



Reminder: SETH and OV

Problem k -SAT:

Given formula in k -CNF with n variables, is it satisfiable?

$$(x_1 \vee \neg x_2 \vee x_3) \wedge (x_2 \vee \neg x_3 \vee x_5) \wedge (\neg x_1 \vee x_4 \vee x_5)$$

Strong Exponential Time Hypothesis: [Impagliazzo, Paturi'01]

$\forall \varepsilon > 0 \exists k$: k -SAT has no $O(2^{(1-\varepsilon)n})$ -time algorithm

SETH

↓ [Williams'05]

Problem Orthogonal Vectors:

Given sets $A, B \subseteq \{0,1\}^d$ of size n ,
are any $a \in A, b \in B$ orthogonal? ($\sum_i a_i \cdot b_i = 0$)

LD-OVH

↓

OVH

OV-Hypothesis: (moderate dimension)

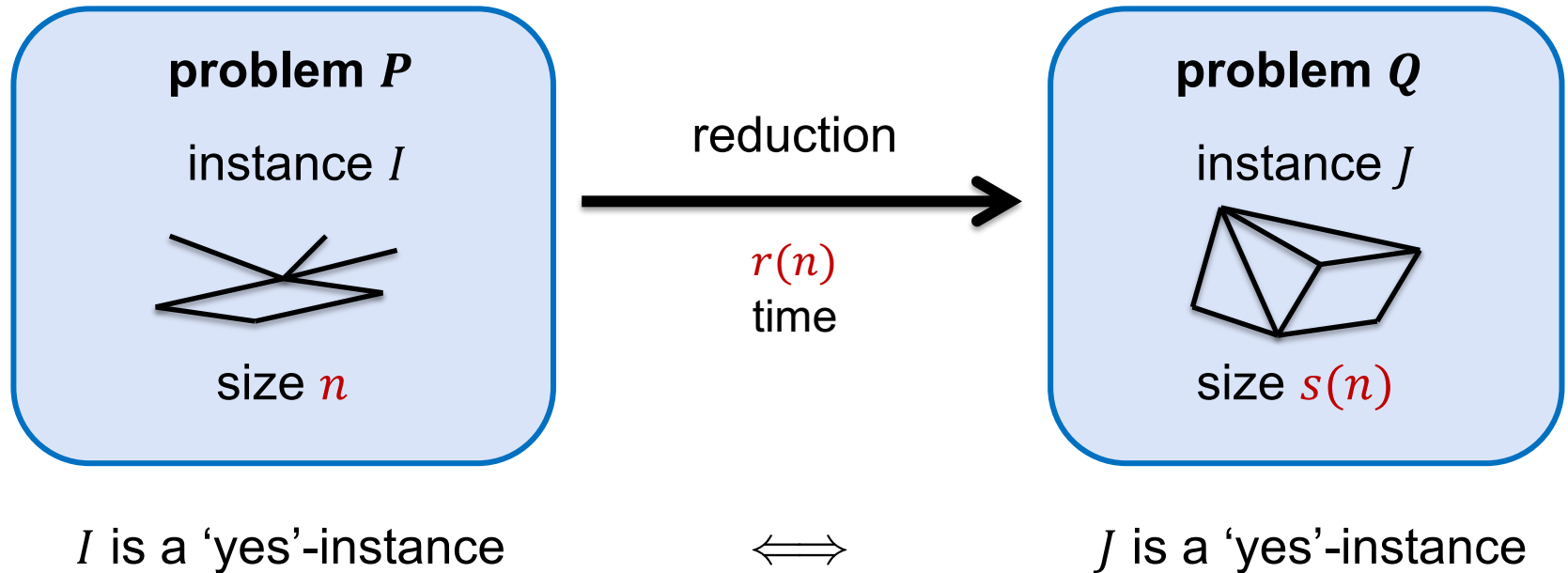
$\forall \varepsilon > 0$: OV has no $O(n^{2-\varepsilon} \cdot \text{poly}(d))$ -time algorithm

LD-OV-Hypothesis: (low dimension)

$\forall \varepsilon > 0 \exists c > 0$: OV in $d = c \log n$ has no $O(n^{2-\varepsilon})$ -time algorithm

Reminder: Fine-Grained Reductions

transfer hardness of one problem to another one by reductions

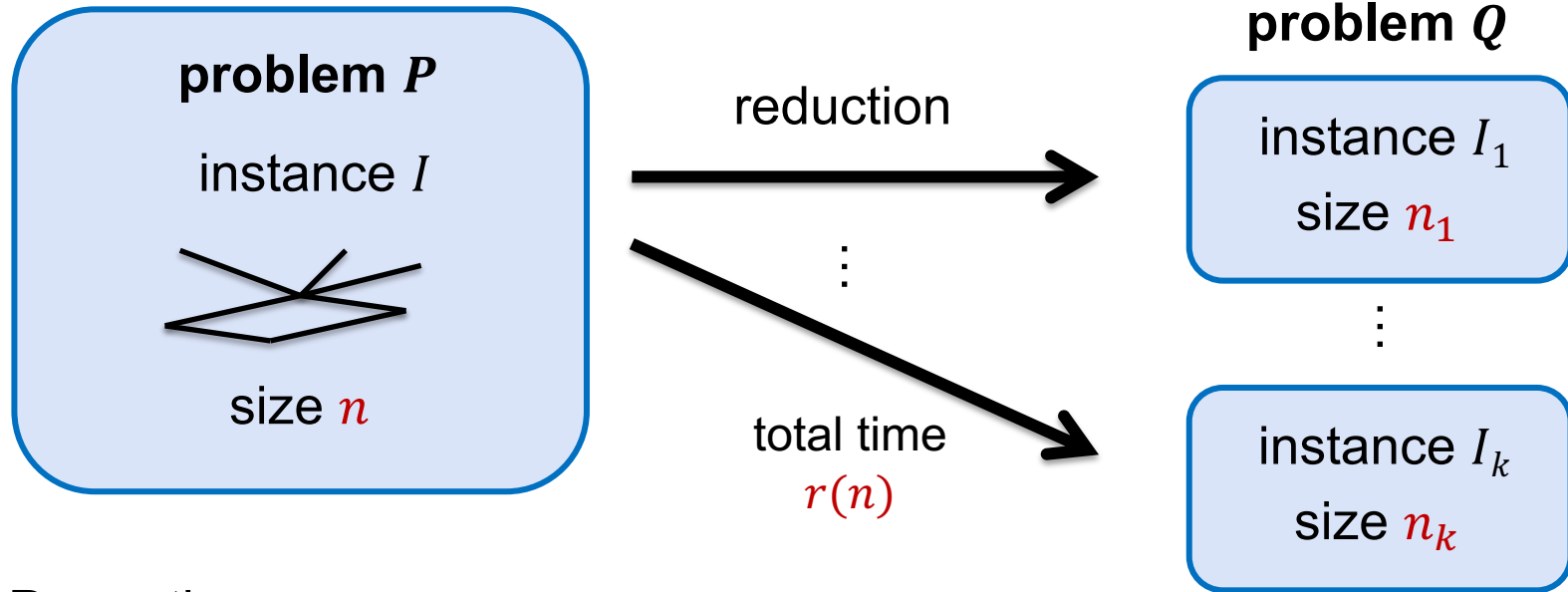


$t(n)$ algorithm for Q implies a $r(n) + t(s(n))$ algorithm for P

if P has no $r(n) + t(s(n))$ algorithm then Q has no $t(n)$ algorithm

Reminder: Fine-Grained Reductions

A **fine-grained reduction** from (P, T) to (Q, T') is an algorithm A for P with **oracle** access to Q s.t.:



Properties:

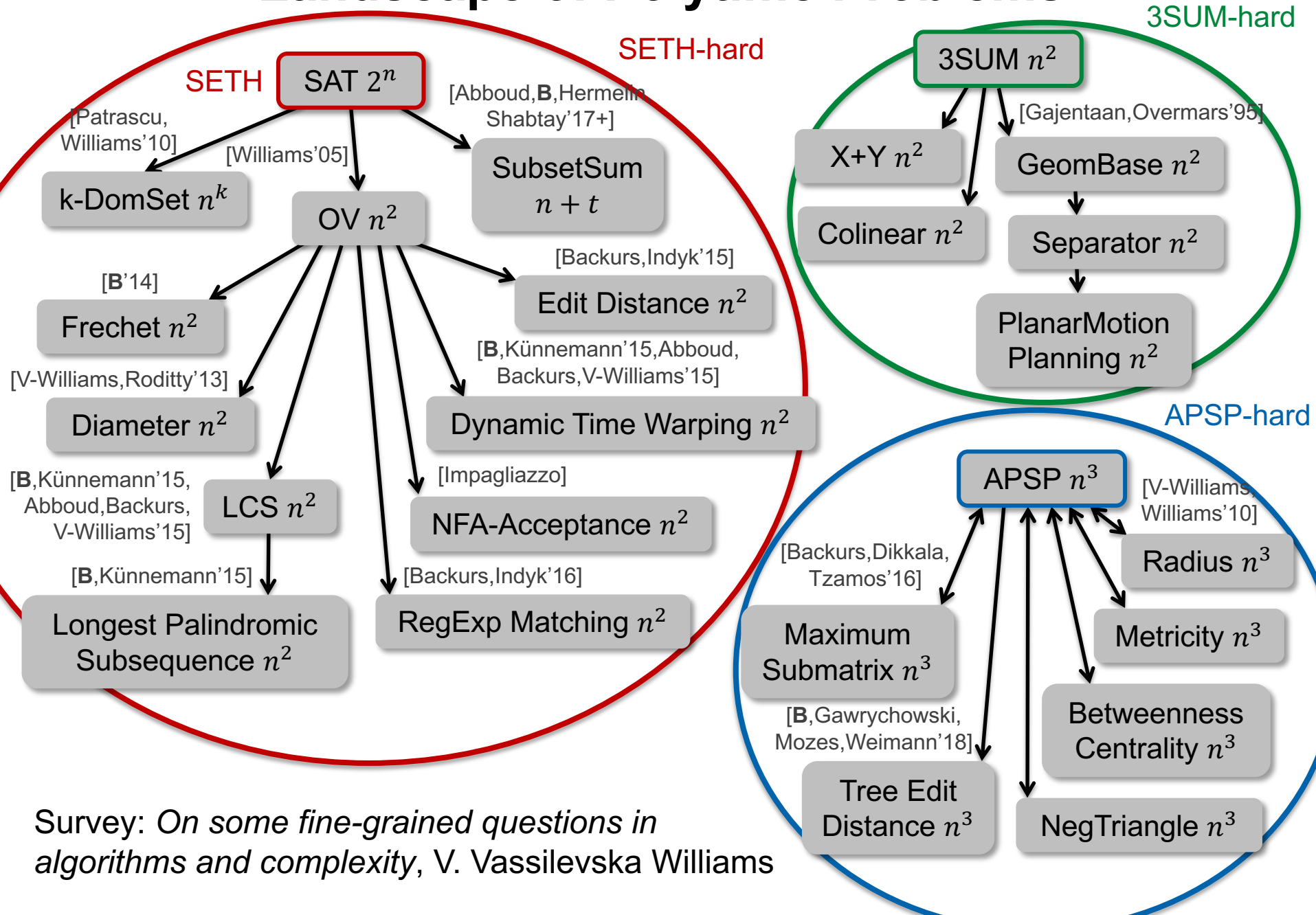
for any instance I , algorithm $A(I)$ correctly solves problem P on I

A runs in time $r(n) = O(T(n)^{1-\gamma})$ for some $\gamma > 0$

for any $\varepsilon > 0$ there is a $\delta > 0$ s.t. $\sum_{i=1}^k T'(n_i)^{1-\varepsilon} \leq T(n)^{1-\delta}$



Landscape of Polytime Problems



Survey: On some fine-grained questions in algorithms and complexity, V. Vassilevska Williams

I. Easy Example

II. Longest Common Subsequence

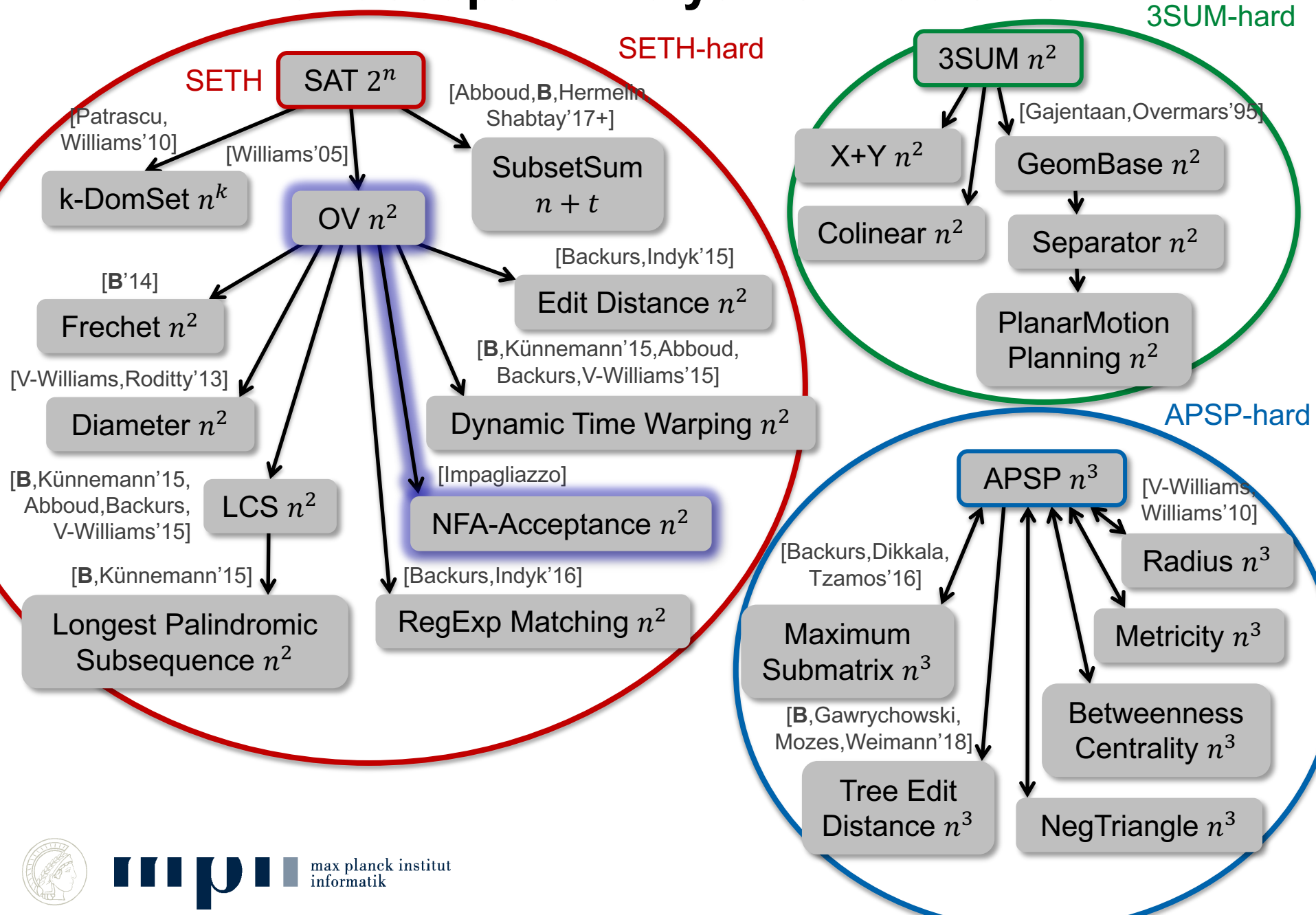
III. Hardness of Approximation

IV. Further Topics

V. Summary

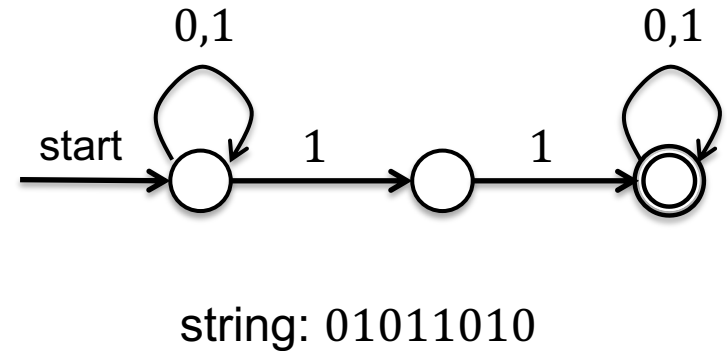


Landscape of Polytime Problems



NFA Acceptance Problem

nondeterministic finite automaton G
accepts input string s if there is a
walk in G from starting state to
some accepting state,
labelled with s



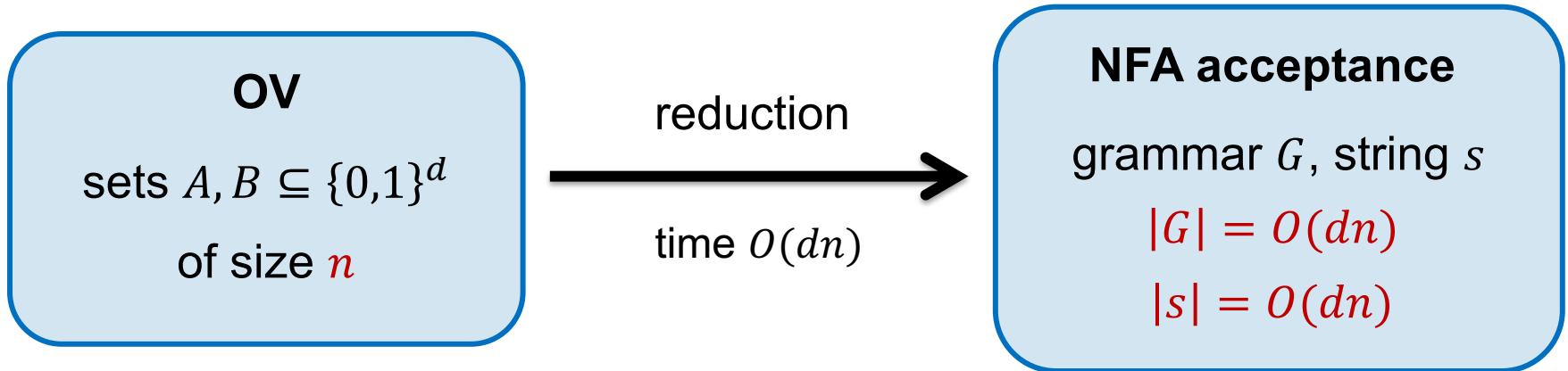
dynamic programming algorithm in time $O(|s||G|)$:

$T[i] :=$ set of states reachable via walks labelled with $s[1..i]$

$T[0] :=$ {starting state}

$T[i] := \{v \mid \exists u \in T[i - 1] \text{ and } \exists \text{ transition } u \rightarrow v \text{ labelled } s[i]\}$

OV-Hardness Result



$O(n^{2-\varepsilon} \text{poly}(d))$ algorithm

\Leftarrow

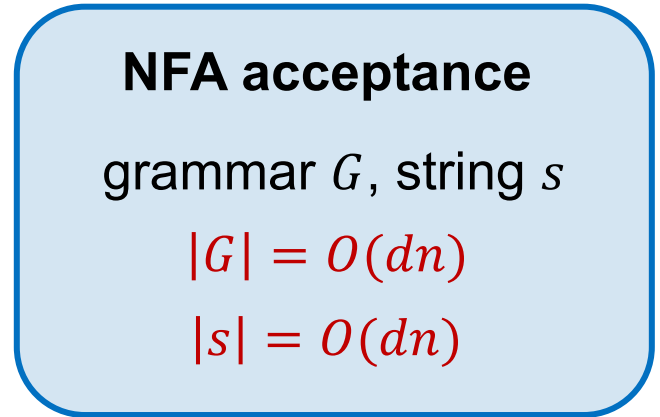
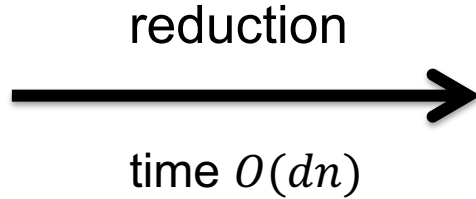
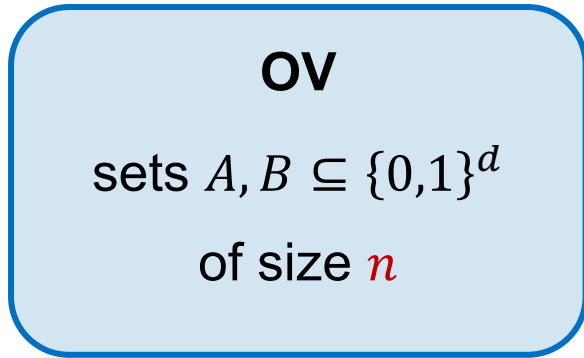
$O((|s| |G|)^{1-\varepsilon})$ algorithm

Thm: NFA acceptance has no $O((|s| |G|)^{1-\varepsilon})$
algorithm unless OVH fails.

[Impagliazzo]



Proof



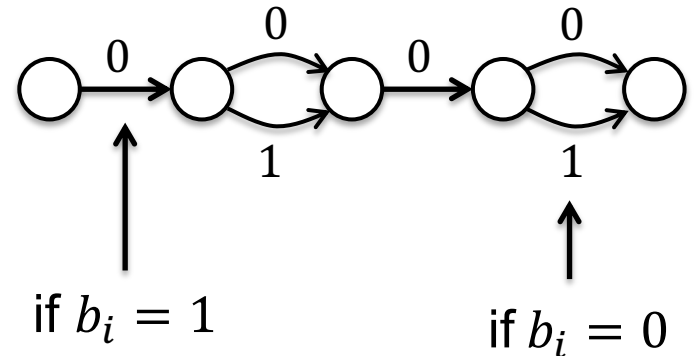
Proof:

for any $a \in A$:
construct string:

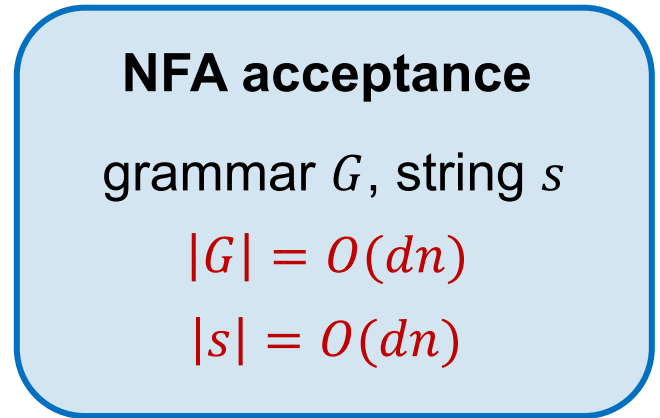
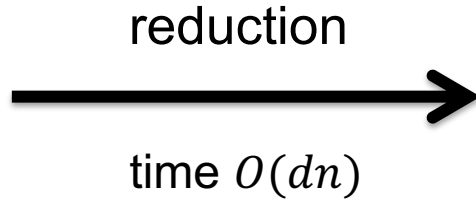
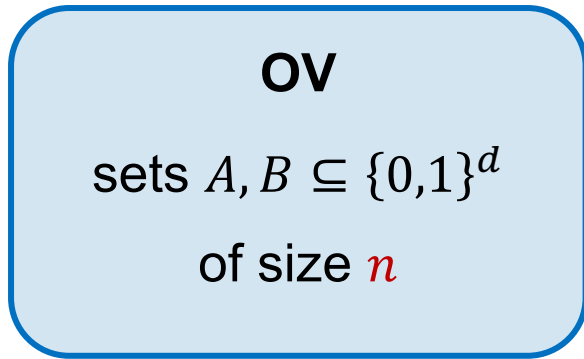
$$\begin{array}{c} 0011 \\ \uparrow \\ = a_1 a_2 \dots a_d \end{array}$$

for any $b \in B$:

construct part of NFA:



Proof



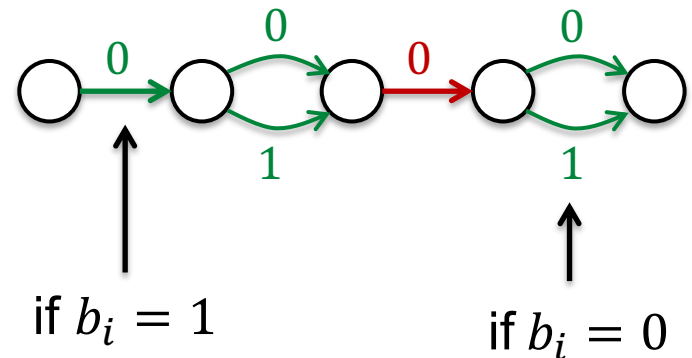
Proof:

for any $a \in A$:
construct string:

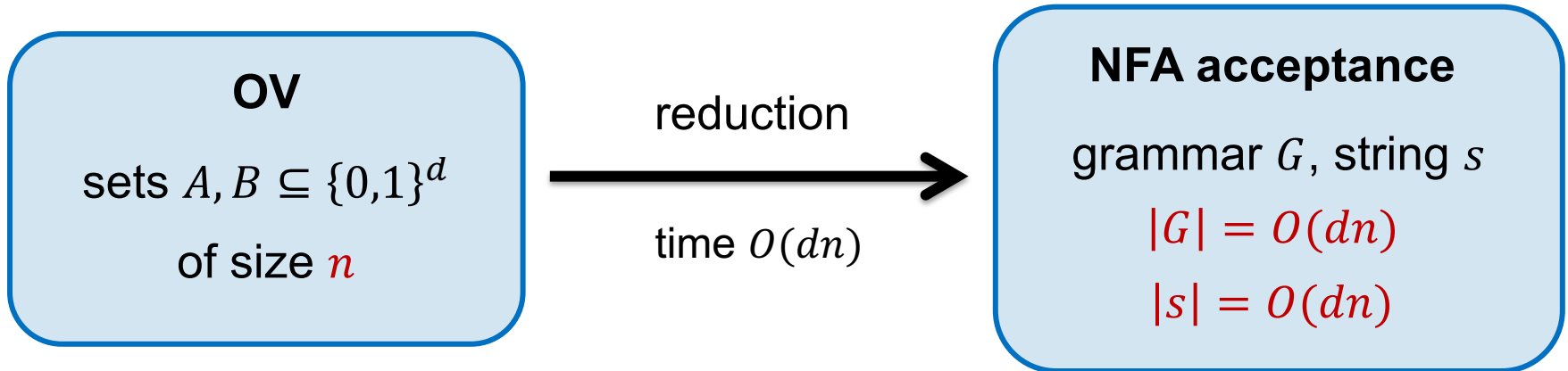
$$\begin{array}{c} 0011 \\ \uparrow \\ = a_1 a_2 \dots a_d \end{array}$$

for any $b \in B$:

construct part of NFA:



OV-Hardness Result



$O(n^{2-\varepsilon} \text{poly}(d))$ algorithm

\Leftarrow

$O((|s| |G|)^{1-\varepsilon})$ algorithm

Thm: NFA acceptance has no $O((|s| |G|)^{1-\varepsilon})$ algorithm unless OVH fails.

[Impagliazzo]



I. Easy Example

II. Longest Common Subsequence

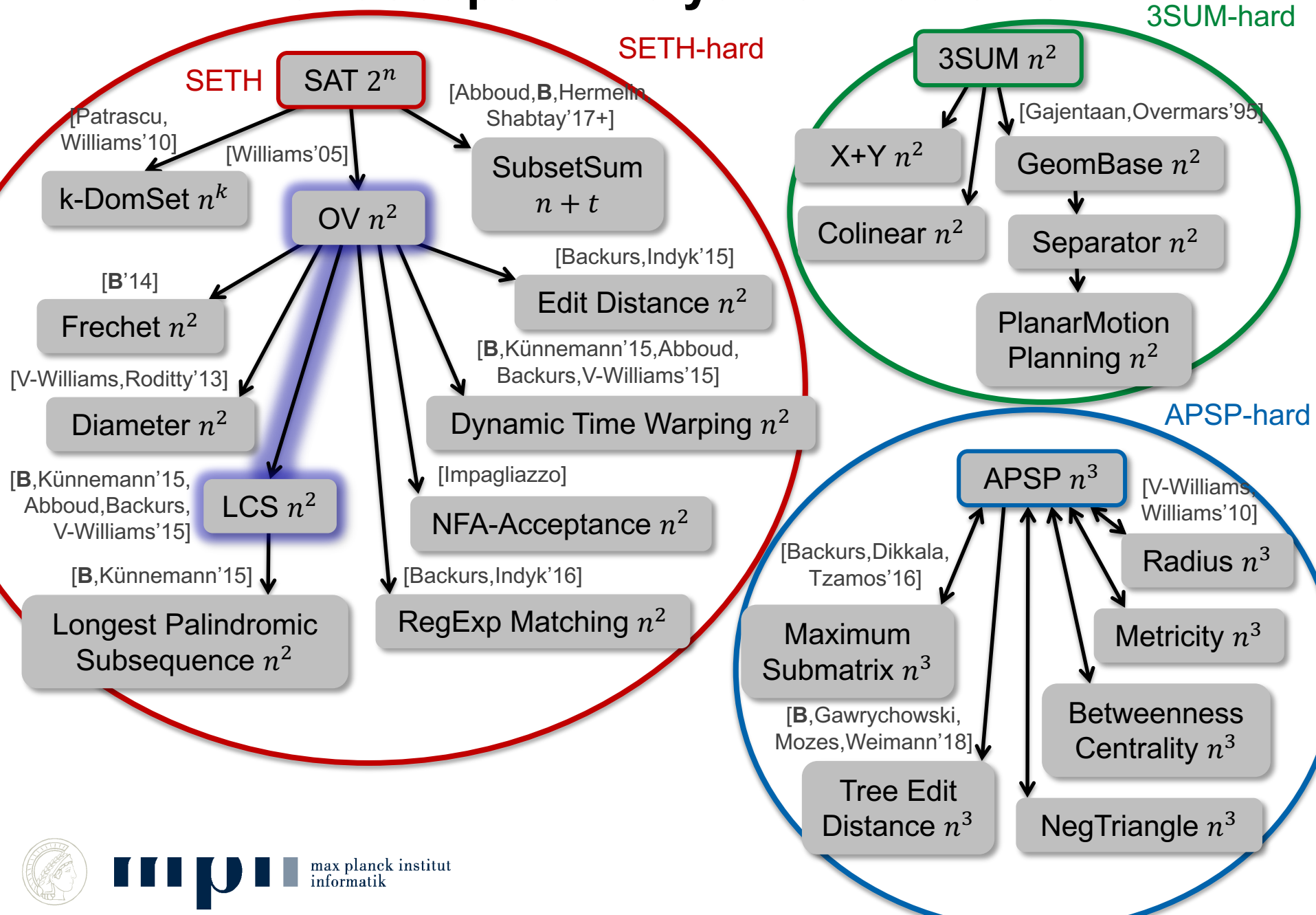
III. Hardness of Approximation

IV. Further Topics

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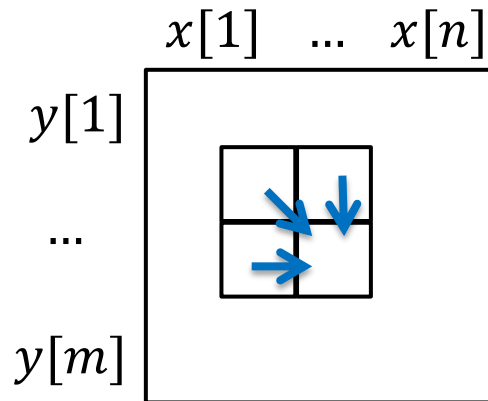
Landscape of Polytime Problems



Longest Common Subsequence (LCS)

given strings x, y of length $n \geq m$, compute longest string z that is a subsequence of both x and y

natural dynamic program $O(n^2)$



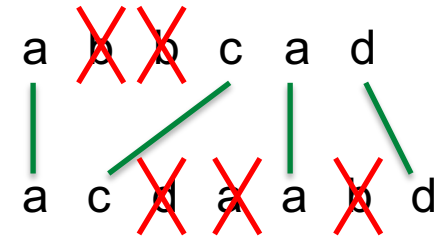
$$T[i, j] = \text{LCS}(x[1..i], y[1..j])$$

$$T[i, j] = \max\{T[i-1, j], T[i, j-1]\}$$

if $x[i] = y[j]$:

$$T[i, j] = \max\{T[i, j], T[i-1, j-1] + 1\}$$

write $\text{LCS}(x, y) = |z|$



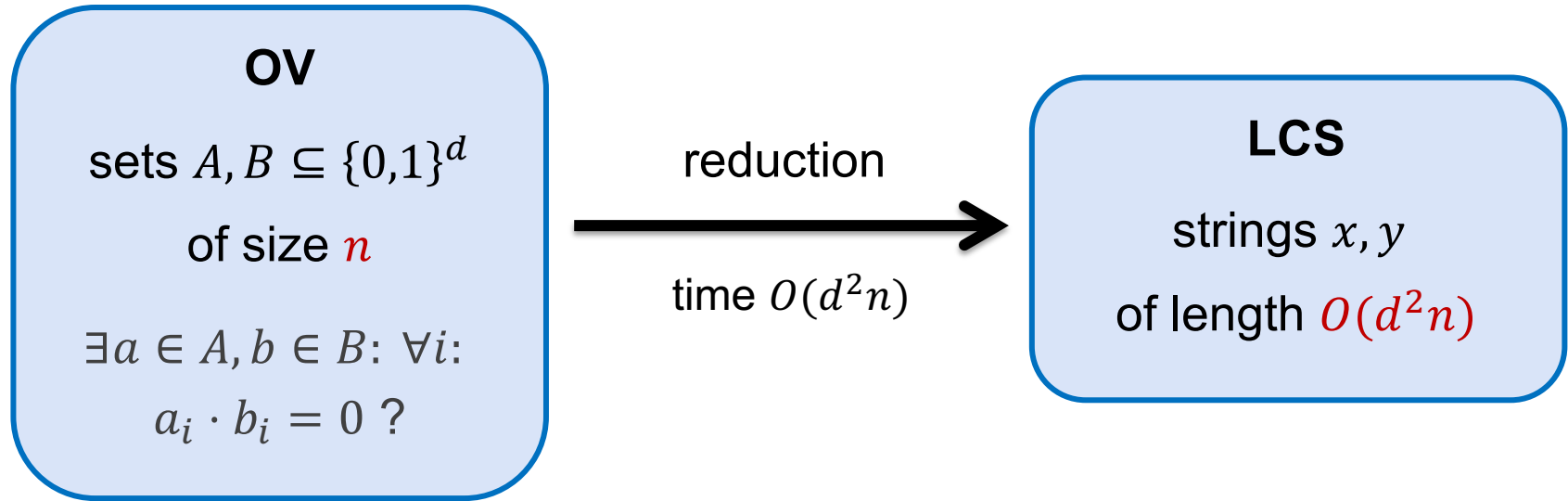
logfactor improvement:

$$O(n^2 / \log^2 n)$$

[Masek, Paterson'80]



OV-Hardness of LCS



$O(n^{2-\varepsilon} \text{poly}(d))$ algorithm

\Leftarrow

$O(n^{2-\varepsilon})$ algorithm

Thm: Longest Common Subsequence
has no $O(n^{2-\varepsilon})$ algorithm unless the OV-Hypothesis fails.

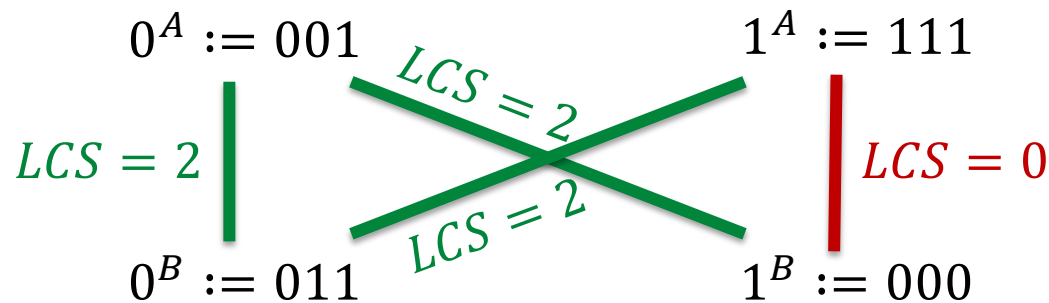
[B.,Künnemann'15
Abboud,Backurs,V-Williams'15]



Proof: Coordinate Gadgets

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

we want to simulate the **coordinates** $\{0,1\}$ and the behavior of $a_i \cdot b_i$



replace a_i by a_i^A and b_i by b_i^B


$LCS(a_i^A, b_i^B)$ can be written as $f(a_i \cdot b_i)$, with $f(0) > f(1)$

Proof: Vector Gadgets

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

we want to simulate **orthogonality** of $a \in A, b \in B$ in the picture: $d = 4$
concatenate a_1^A, \dots, a_d^A , padded with a new symbol 2

length $4d$

$$VG(a) := a_1^A \ 2 \ \dots \ 2 \ a_2^A \ 2 \ \dots \ 2 \ a_3^A \ 2 \ \dots \ 2 \ a_4^A$$

$$VG(b) := b_1^B \ 2 \ \dots \ 2 \ b_2^B \ 2 \ \dots \ 2 \ b_3^B \ 2 \ \dots \ 2 \ b_4^B$$

- no LCS matches symbols in a_i^A with symbols in b_j^B where $i \neq j$

Proof: Vector Gadgets

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
 Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

we want to simulate **orthogonality** of $a \in A, b \in B$

concatenate a_1^A, \dots, a_d^A , padded with a new symbol 2

length $4d$

$$\begin{aligned}
 VG(a) &:= a_1^A \ 2 \dots 2 \ a_2^A \ 2 \dots 2 \ a_3^A \ 2 \dots 2 \ a_4^A \\
 VG(b) &:= b_1^B \ 2 \dots 2 \ b_2^B \ 2 \dots 2 \ b_3^B \ 2 \dots 2 \ b_4^B
 \end{aligned}$$

- no LCS matches symbols in a_i^A with symbols in b_j^B where $i \neq j$
 lose one block of 2's
 cannot make up for it with symbols 0/1 since there are too few of them




Proof: Vector Gadgets

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

we want to simulate **orthogonality** of $a \in A, b \in B$

concatenate a_1^A, \dots, a_d^A , padded with a new symbol 2

$$VG(a) := a_1^A \ 2 \ \dots \ 2 \ a_2^A \ 2 \ \dots \ 2 \ a_3^A \ 2 \ \dots \ 2 \ a_4^A$$

$$VG(b) := b_1^B \ 2 \ \dots \ 2 \ b_2^B \ 2 \ \dots \ 2 \ b_3^B \ 2 \ \dots \ 2 \ b_4^B$$

- no LCS matches symbols in a_i^A with symbols in b_j^B where $i \neq j$
- some LCS matches all 2's




Proof: Vector Gadgets

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
 Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

we want to simulate **orthogonality** of $a \in A, b \in B$

concatenate a_1^A, \dots, a_d^A , padded with a new symbol 2

$$VG(a) := a_1^A \ 2 \ \dots \ 2 \ a_2^A \ 2 \ \dots \ 2 \ a_3^A \ 2 \ \dots \ 2 \ a_4^A$$


$$VG(b) := b_1^B \ 2 \ \dots \ 2 \ b_2^B \ 2 \ \dots \ 2 \ b_3^B \ 2 \ \dots \ 2 \ b_4^B$$

$$- LCS(VG(a), VG(b)) = (d-1)4d + \sum_{i=1}^d LCS(a_i^A, b_i^B) = f(a_i \cdot b_i)$$

#2's

$$LCS(VG(a), VG(b)) = C + 2 \quad \text{if } a \perp b$$

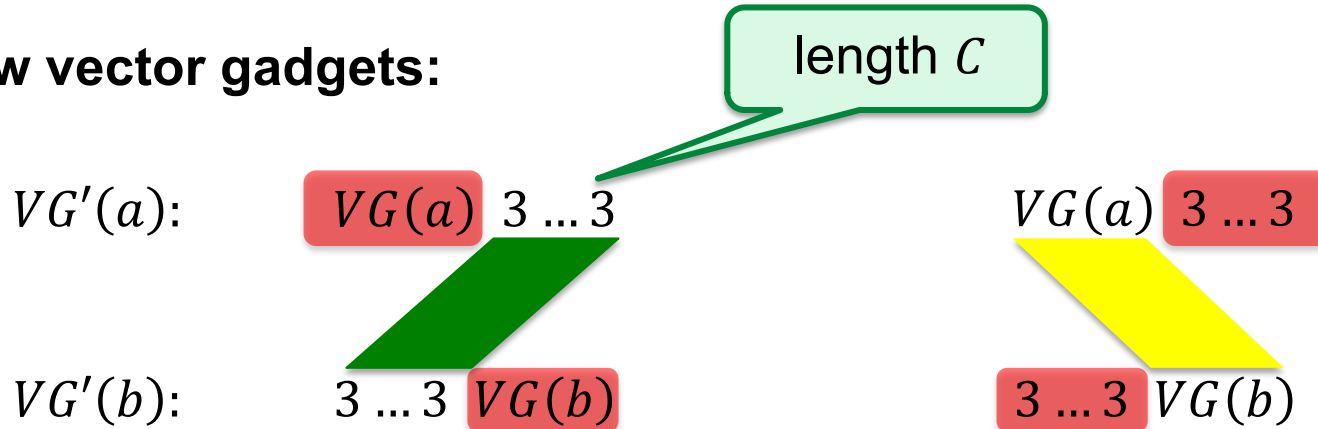
$$LCS(VG(a), VG(b)) \leq C \quad \text{otherwise}$$

for some constant C

Proof: Normalized Vectors Gadgets

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
 Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

new vector gadgets:



$$LCS(VG'(a), VG'(b)) = \max\{LCS(VG(a), VG(b)), C\}$$

$$LCS(VG'(a), VG'(b)) = \begin{cases} C + 2 & \text{if } a \perp b \\ C & \text{otherwise} \end{cases}$$

write VG for VG'

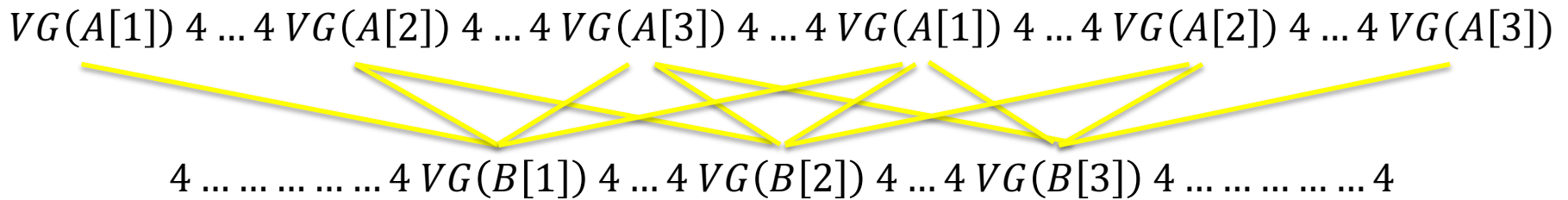


Proof: OR-Gadget

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
 Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

fresh symbol 4, want to construct:

in the picture: $n = 3$



length $100d^2$

length $100d^2 \cdot 2n$

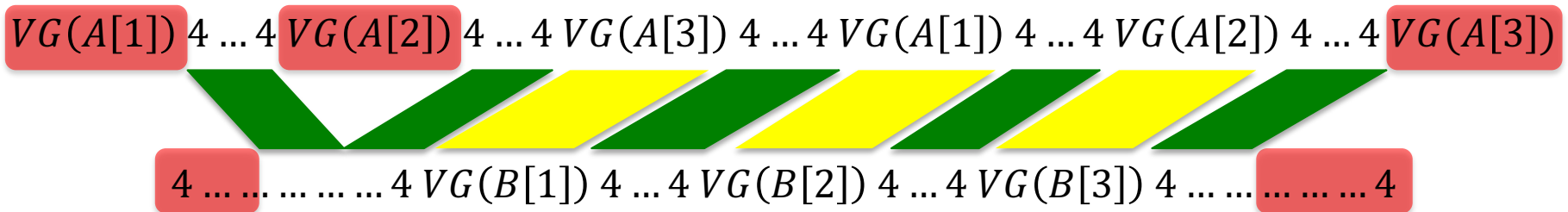


Proof: OR-Gadget

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
 Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

fresh symbol 4, want to construct:

in the picture: $n = 3$



can align $VG(B[j])$ with $VG(A[\Delta + j \bmod n])$ for any offset Δ

$$LCS \geq \underbrace{(2n - 1)100d^2}_{\text{\#4's in upper string}} + \max_{\Delta} \sum_{j=1}^n LCS(VG(A[\Delta + j \bmod n]), VG(B[j]))$$

#4's in upper string

maximize over offset

need normalization!

If there is an orthogonal pair, some offset Δ aligns this pair, and we get

$$LCS \geq (2n - 1)100d^2 + nC + 2$$

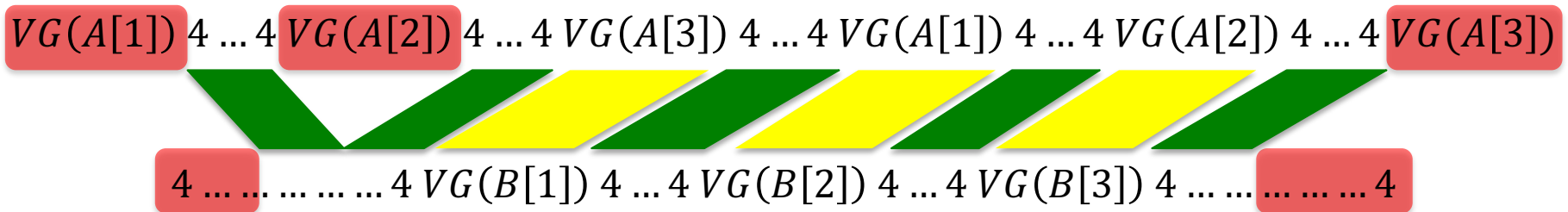


Proof: OR-Gadget

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
 Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

fresh symbol 4, want to construct:

in the picture: $n = 3$



if an orthogonal pair exists then $LCS \geq (2n - 1)100d^2 + nC + 2$

Claim: otherwise: $LCS \leq (2n - 1)100d^2 + nC$

this finishes the proof:

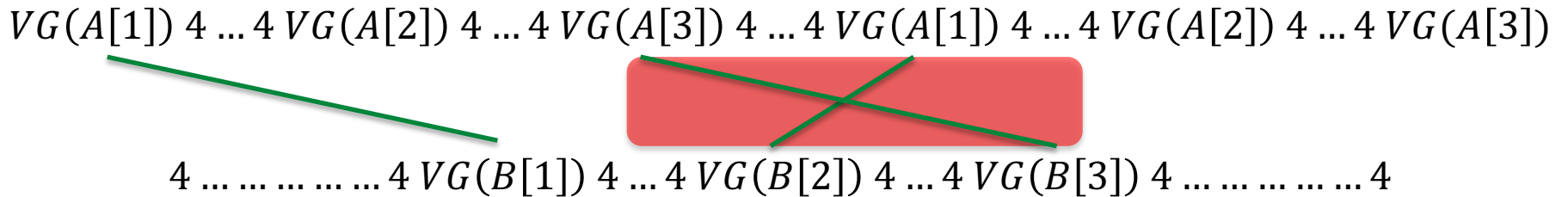
- equivalent to OV instance
- length $O(d^2n)$



Proof of Claim

OV: Given $A, B \subseteq \{0,1\}^d$ of size n each
 Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

Claim: if no orthogonal pair exists: $LCS \leq (2n - 1)100d^2 + nC$



consider how an LCS matches the $VG(B[j])$

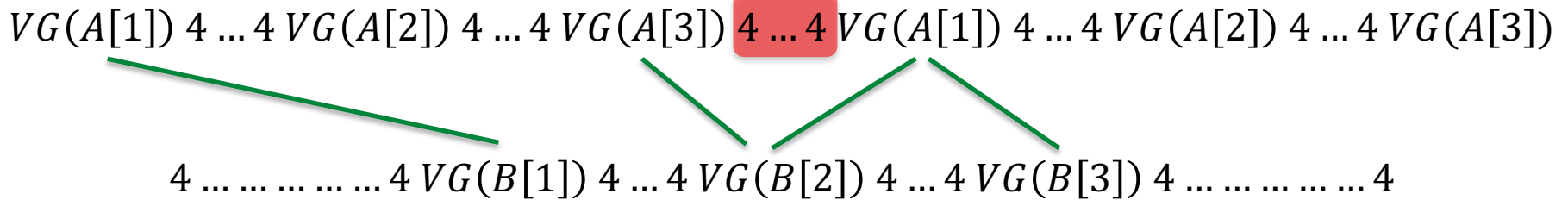
- no crossings

Proof of Claim

Given $A, B \subseteq \{0,1\}^d$ of size n each

Are there $a \in A, b \in B$ such that $\forall i: a_i \cdot b_i = 0$

Claim: if no orthogonal pair exists: $LCS \leq (2n - 1)100d^2 + nC$



non-orthogonal

$LCS \leq (2n - 1)100d^2 + \sum_{j=1}^n$

#4's in upper string

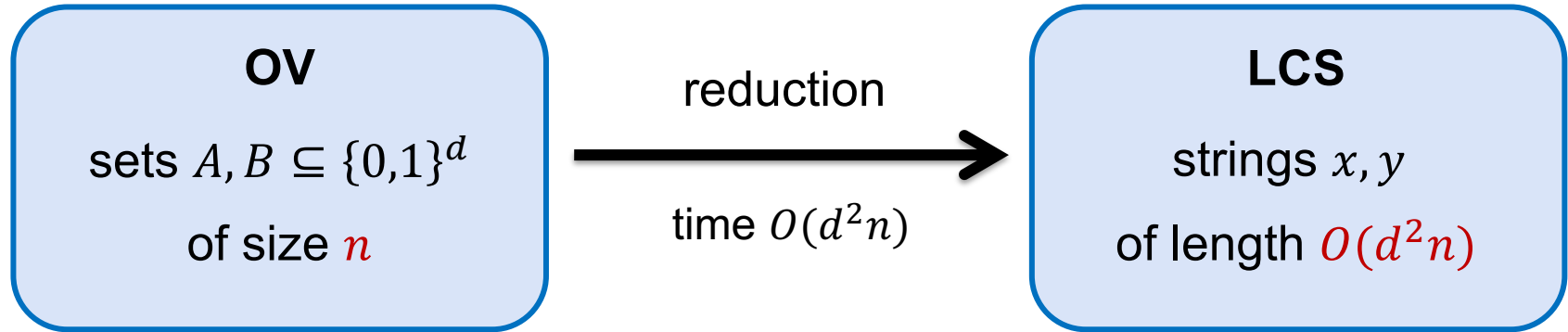
$\begin{cases} 0 & \text{if } VG(B[j]) \text{ is not matched} \\ C & \text{if } VG(B[j]) \text{ is matched to one} \\ |VG(B[j])| - |4 \dots 4| & \text{if } VG(B[j]) \text{ is matched to } > 1 \end{cases}$

could match VG completely, but loose many 4's

≤ 0



OV-Hardness of LCS



$O(n^{2-\varepsilon} \text{poly}(d))$ algorithm

\Leftarrow

$O(n^{2-\varepsilon})$ algorithm

Thm: Longest Common Subsequence
has no $O(n^{2-\varepsilon})$ algorithm unless the OV-Hypothesis fails.

[B., Künnemann'15
Abboud, Backurs, V-Williams'15]



Extensions

similar problems: edit distance, dynamic time warping, ...

[Backurs,Indyk'15] [B.,Künnemann'15,Abboud,Backurs,V-Williams'15]

harder hardness: reductions from Formula-SAT / branching programs

[Abboud,Hansen,V-Williams,Williams'16] [Abboud,B.'18]

LCS-hard problems: longest palindromic subseq., longest tandem subseq.

[B.,Künnemann'15]

alphabet size:

[B.,Künnemann'15]

LCS and edit distance are hard on *binary* strings, i.e., alphabet $\{0,1\}$

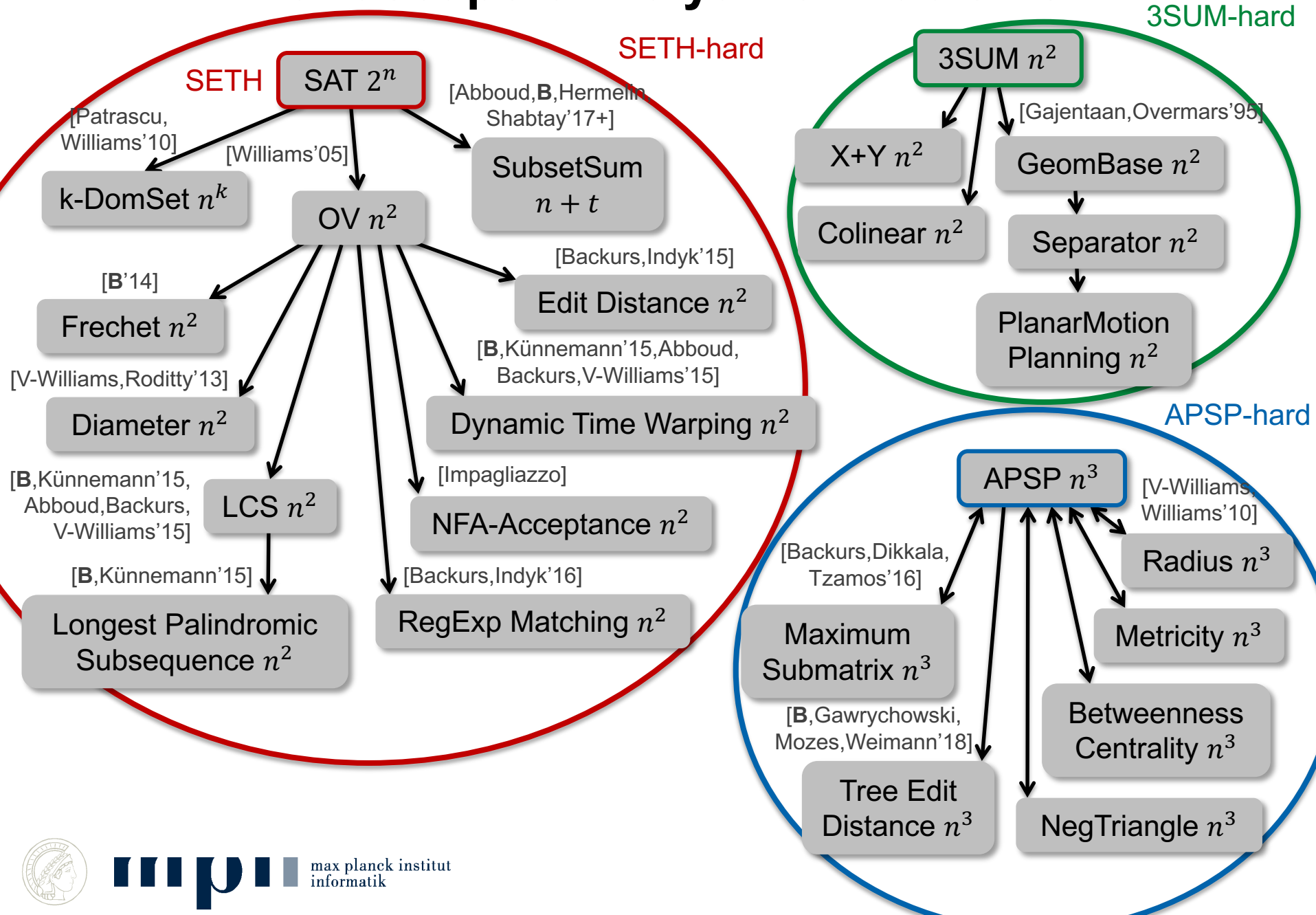
k-LCS: LCS of k strings takes time $\Omega(n^{k-\varepsilon})$

[Abboud,Backurs,V-Williams'15]

Based on k -OV



Landscape of Polytime Problems



I. Easy Example

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Hardness of Approximation in P

**„deterministic subquadratic-time approximation algorithms
imply circuit lower bounds“**

[Abboud,Backurs ITCS'17
Abboud,Rubinstein ITCS'18]

e.g. Longest Common Subsequence

„subquadratic-time approximation algorithms violate SETH“

[Abboud,Rubinstein,Williams FOCS'17
Rubinstein STOC'18
Chen CCC'18

e.g. Max-Inner-Product

Chen,Goldwasser,Lyu,Rothblum,Rubinstein Arxiv'18]

„FPT-approximation algorithms violate SETH“

[Chalermsook,Cygan,Kortsarz,Laekhanukit,Manurangsi,Nanongkai,Trevisan FOCS'17
Karthik,Laekhanukit,Manurangsi STOC'18]

Dominating Set



Hardness of Approximation in P

Problem Max-Inner-Product:

Given sets $A, B \subseteq \{0,1\}^d$ of size n ,
compute $\max_{a \in A, b \in B} \langle a, b \rangle$

„optimization version of OV“

Is in time $O(n^2 d)$, but not in time $O(n^{2-\varepsilon} \cdot \text{poly}(d))$ assuming OV-H

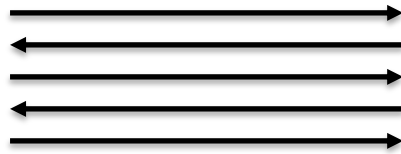
Thm: Let $\log n \ll d \leq n^{o(1)}$.

[Chen CCC'18]

- 1) $\forall \delta > 0 \exists \varepsilon > 0$: A $(d/\log n)^\delta$ -multiplicative approximation to Max-IP can be computed in time $O(n^{2-\varepsilon} \cdot \text{poly}(d))$
- 2) If $\exists \varepsilon > 0 \forall \delta > 0$: a $(d/\log n)^\delta$ -multiplicative approximation to Max-IP can be computed in time $O(n^{2-\varepsilon} \cdot \text{poly}(d))$, then LD-OVH fails



Tool: Communication Complexity



Input: $x \in \{0,1\}^d$

Input: $y \in \{0,1\}^d$

want to compute $f(x, y)$

with as few **communication**
as possible

e.g. $f = \text{Disj}$:

$$\text{Disj}(x, y) = \begin{cases} 1 & \text{if } \langle x, y \rangle = 0 \\ 0 & \text{otherwise} \end{cases}$$

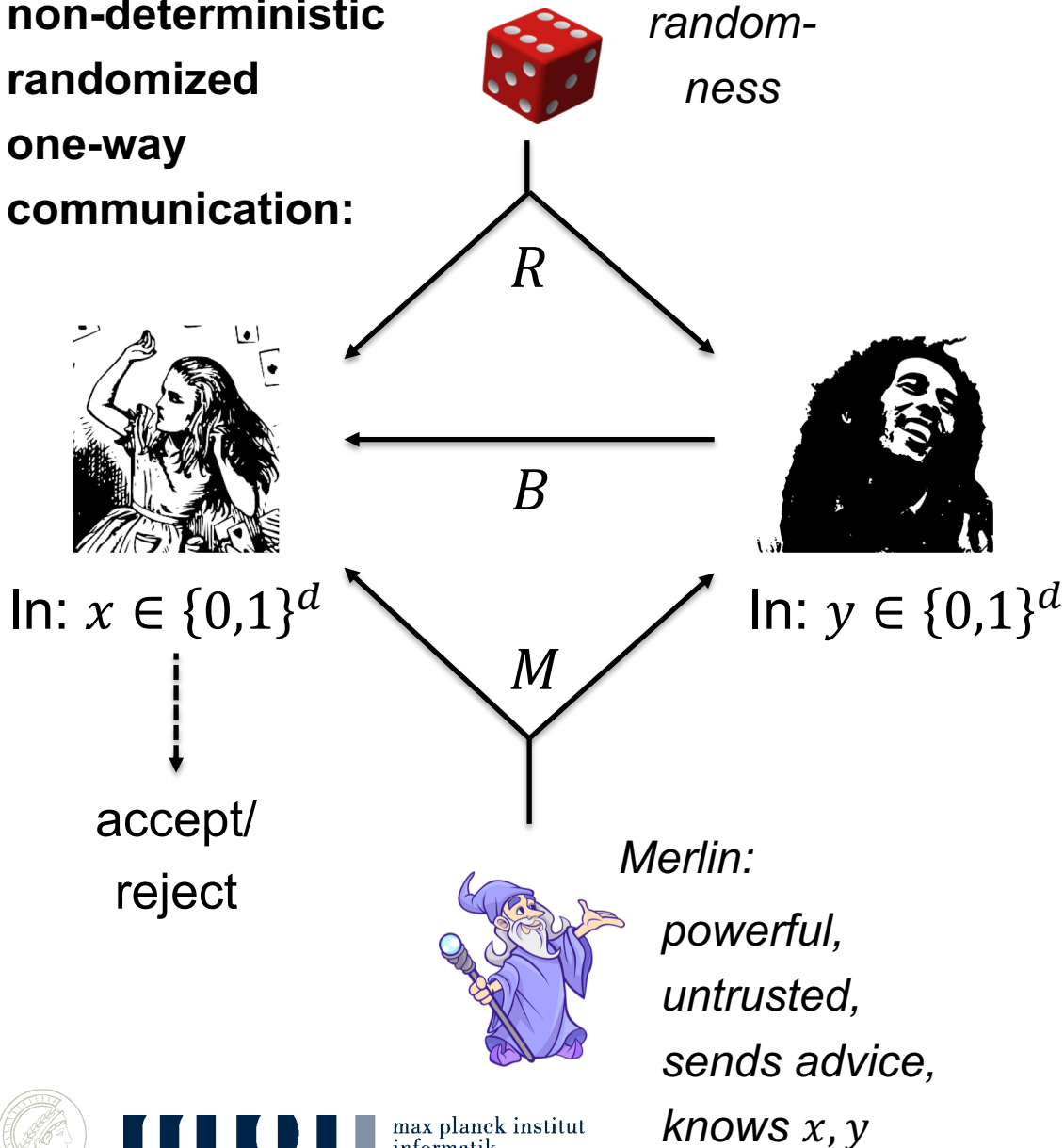
deterministic communication
complexity of Disj is $\Theta(d)$

randomized communication
complexity of Disj is $\Theta(d)$



Communication Complexity Setup

non-deterministic
randomized
one-way
communication:



Messages M, R, B
of bitlength m, r, b

We require:

Efficiency: Alice and Bob
are efficiently computable

Completeness: If $f(x, y) = 1$:
 $\exists M: \Pr_R[\text{Alice accepts}] = 1$

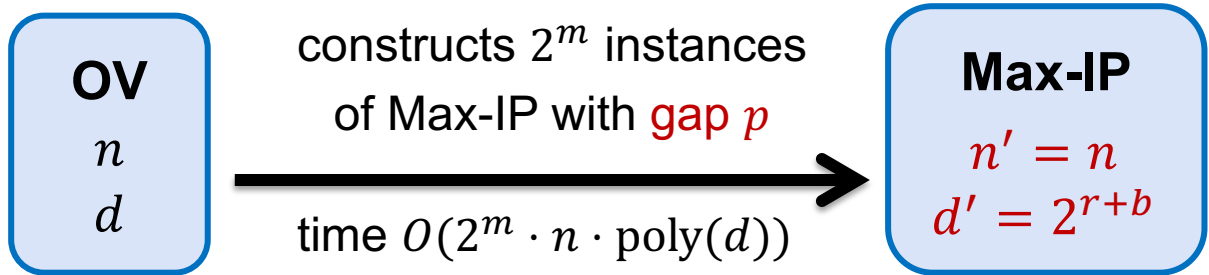
Soundness: If $f(x, y) = 0$:
 $\forall M: \Pr_R[\text{Alice accepts}] \leq p$

Then we say that f has an
 (m, r, b, p) -efficient protocol



From CC to Reductions

Any (m, r, b, p) -efficient protocol for *Disj* yields a reduction:



YES-instance

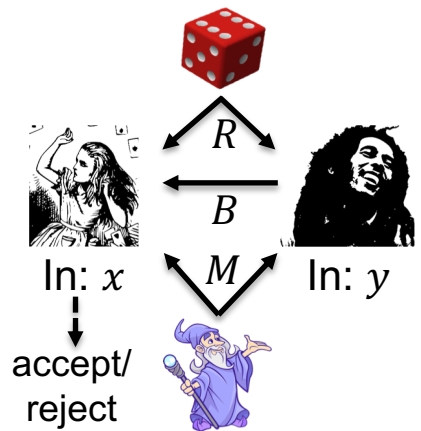
\Rightarrow

Max-IP $\geq 2^r$
for some instance

NO-instance

\Rightarrow

Max-IP $\leq p \cdot 2^r$
for all instances



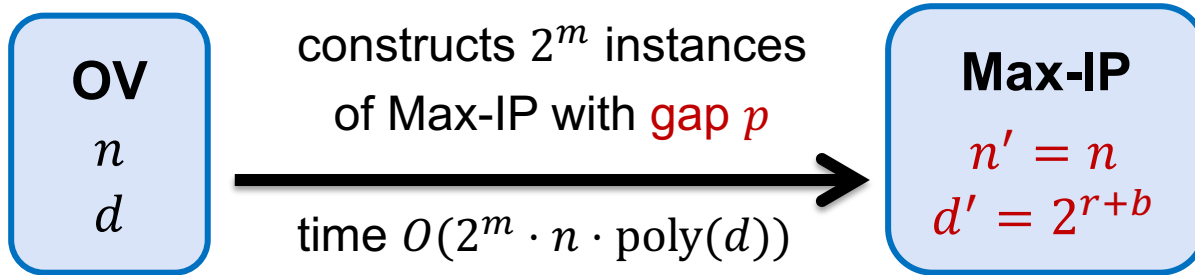
Alice and Bob efficient

Complete: If $f(x, y) = 1$:
 $\exists M: \Pr_R[\text{Alice accepts}] = 1$

Sound: If $f(x, y) = 0$:
 $\forall M: \Pr_R[\text{Alice accepts}] \leq p$

From CC to Reductions

Any (m, r, b, p) -efficient protocol for $Disj$ yields a reduction:



Proof: Given OV instance $X, Y \subseteq \{0,1\}^d$

Iterate over all advice strings $M \in [2^m]$:

Construct Max-IP instance X^M, Y^M :

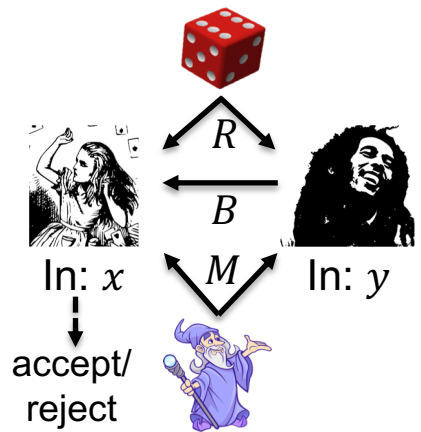
$\forall x \in X: \forall R \in [2^r]:$ create $x^{M,R} \in \{0,1\}^{2^b}$ with

$x_B^{M,R}$ = whether Alice accepts on input x , advice M , randomness R and Bob's message B

$\forall y \in Y: \forall R \in [2^r]:$ create $y^{M,R} \in \{0,1\}^{2^b}$ with

$y_B^{M,R}$ = whether Bob sends message B on input y , advice M and randomness R

Then $\langle x^{M,R}, y^{M,R} \rangle$ = whether Alice accepts on inputs x, y , advice M , randomness R



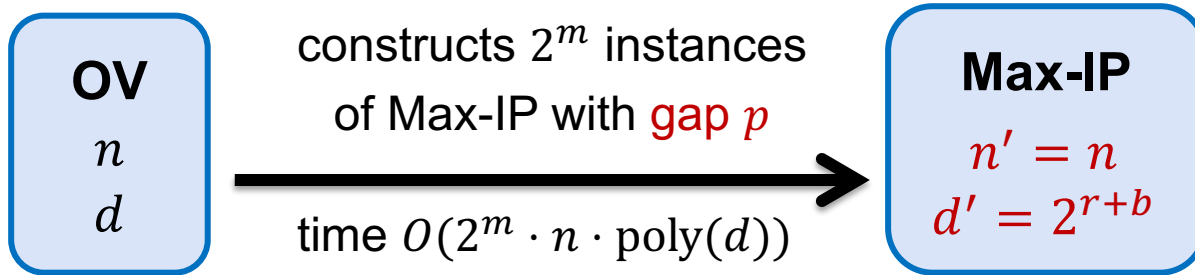
Alice and Bob efficient

Complete: If $f(x, y) = 1$:
 $\exists M: \Pr_R[\text{Alice accepts}] = 1$

Sound: If $f(x, y) = 0$:
 $\forall M: \Pr_R[\text{Alice accepts}] \leq p$

From CC to Reductions

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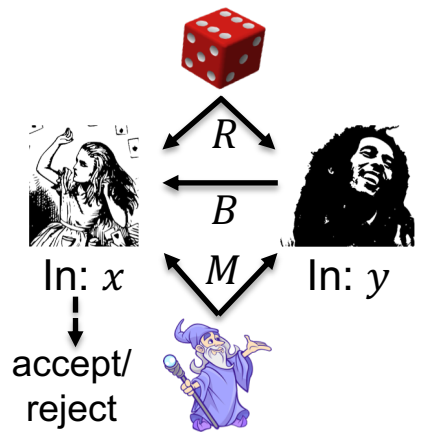
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$\forall y \in Y$: create y^M by concatenating all $y^{M,R}$

Then $\langle x^M, y^M \rangle = 2^r \cdot \Pr_R[\text{Alice accepts on inputs } x, y \text{ and advice } M]$

Then $\langle x^{M,R}, y^{M,R} \rangle =$ whether Alice accepts on inputs x, y , advice M , randomness R



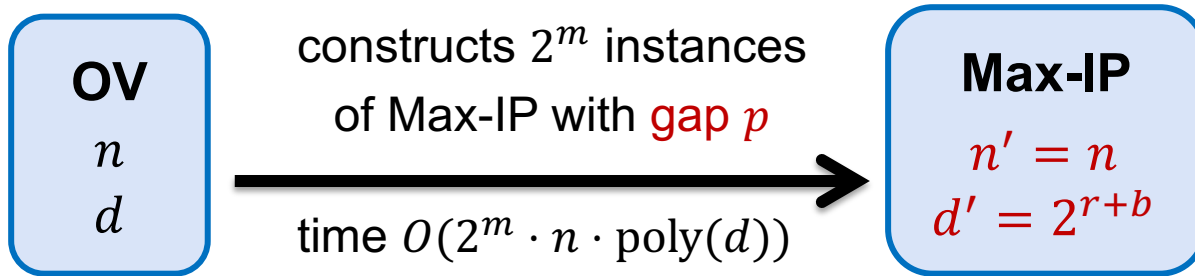
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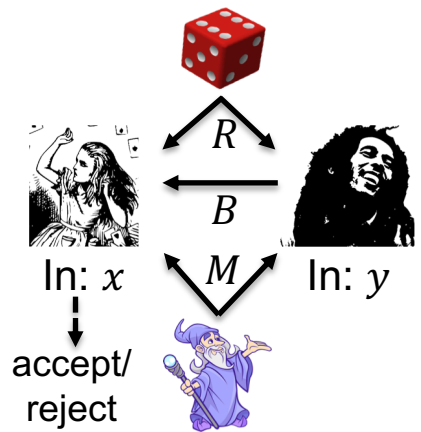
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Then $\langle x^M, y^M \rangle = 2^r \cdot \Pr_R[\text{Alice accepts on inputs } x, y \text{ and advice } M]$

If $\exists x \in X, y \in Y: \langle x, y \rangle = 0$ then: $\exists M: \langle x^M, y^M \rangle = 2^r$

Otherwise: $\forall x \in X, y \in Y: \forall M: \langle x^M, y^M \rangle \leq p \cdot 2^r$



Alice and Bob efficient

Complete: If $f(x, y) = 1$:
 $\exists M: \Pr_R[\text{Alice accepts}] = 1$

Sound: If $f(x, y) = 0$:
 $\forall M: \Pr_R[\text{Alice accepts}] \leq p$

From CC to Reductions

Best known protocol for Disj:

For any d, α and $p < 1/2$:

[Chen CCC'18]

$(\frac{d}{\alpha}, \log_2 d + O(\log \frac{1}{p}), \text{poly}(\alpha) \cdot \log \frac{1}{p}, p)$ -efficient protocol

Complicated: algebraic codes, expander mixing lemma, ...

Easier:

[Aaronson, Wigderson'09]

$(m, r, b, p) = (o(d), o(d), o(d), \frac{1}{2})$ -efficient protocol

Implies reduction:

OV
 n
 $d = c \log n$

constructs $n^{o(1)}$ instances
of Max-IP with **gap 1/2**

time $O(n^{1+o(1)} \cdot \text{poly}(d))$

Max-IP
 $n' = n$
 $d' = n^{o(1)}$

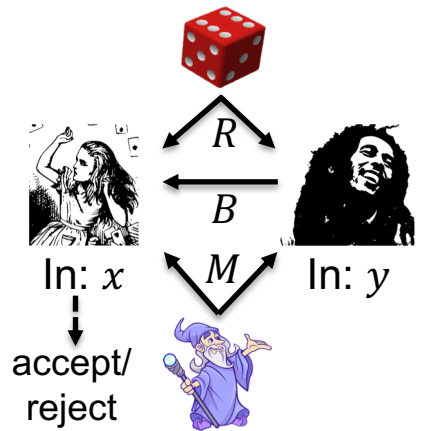
$$2^m, 2^{r+b} = n^{o(1)}$$

[Abboud, Rubinfeld, Williams'17]



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So a **1.99**-approximation of Max-IP in dimension
 $d = n^{o(1)}$ in time $O(n^{2-\epsilon})$ violates LD-OVH



Alice and Bob efficient

Complete: If $f(x, y) = 1$:
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The Disj-Protocol

[Aaronson,
Wigderson'09]

$(m, r, b, p) = (\tilde{O}(\sqrt{d}), O(\log d), \tilde{O}(\sqrt{d}), \frac{1}{2})$ -efficient protocol

Proof: Fix prime $p \in (4d, 8d]$, $T := \sqrt{d}$, assume that $T \in \mathbb{N}$

Given x, y compute polynomials of degree $< d/T$ over \mathbb{Z}_p :

$$\psi_{x,t}(Z) \text{ s.t. } \psi_{x,t}(i) = x_{i \cdot T + t} \text{ for } 0 \leq i < d/T$$

$$\psi_{y,t}(Z) \text{ s.t. } \psi_{y,t}(i) = y_{i \cdot T + t} \text{ for } 0 \leq i < d/T$$

Define $\Phi(Z) := \sum_{t=1}^T \psi_{x,t}(Z) \cdot \psi_{y,t}(Z)$ of degree $< 2d/T$

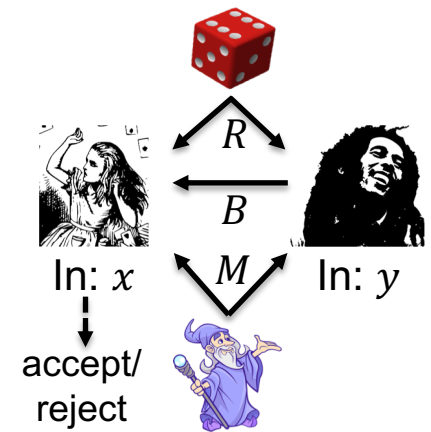
Protocol:

randomness: $R \in \mathbb{Z}_p$ uniformly at random

Bob: sends $\psi_{y,1}(R), \dots, \psi_{y,T}(R)$

Merlin: sends polynomial $P(Z)$ of degree $< 2d/T$ over \mathbb{Z}_p
„good proof“: $P(Z) = \Phi(Z)$

Alice: accepts iff $P(i) = 0$ for all $0 \leq i < d/T$ and



Alice and Bob efficient

Complete: If $f(x, y) = 1$:
 $\exists M: \Pr_R[\text{Alice accepts}] = 1$

Sound: If $f(x, y) = 0$:
 $\forall M: \Pr_R[\text{Alice accepts}] \leq p$

x, y disjoint iff
 $\Phi(i) := 0$ for
all $0 \leq i < d/T$



$$P(R) = \sum_{t=1}^T \psi_{x,t}(R) \cdot \psi_{y,t}(R)$$

Hardness of Approximation in P

Problem Max-Inner-Product:

Given sets $A, B \subseteq \{0,1\}^d$ of size n ,
compute $\max_{a \in A, b \in B} \langle a, b \rangle$

„optimization version of OV“

Is in time $O(n^2 d)$, but not in time $O(n^{2-\varepsilon} \cdot \text{poly}(d))$ assuming OV-H

Thm: Let $\log n \ll d \leq n^{o(1)}$.

[Chen CCC'18]

1) $\forall \delta > 0 \exists \varepsilon > 0$: A $(d/\log n)^\delta$ -multiplicative approximation to Max-IP can be computed in time $O(n^{2-\varepsilon} \cdot \text{poly}(d))$

2) If $\exists \varepsilon > 0 \forall \delta > 0$: a $(d/\log n)^\delta$ -multiplicative approximation to Max-IP can be computed in time $O(n^{2-\varepsilon} \cdot \text{poly}(d))$, then LD-OVH fails



I. Easy Example

II. Longest Common Subsequence

III. Hardness of Approximation

IV. Further Topics

V. Summary



Multivariate Algorithms for LCS

parameters for LCS:

[B.,Künnemann'18]

$n = x $.. length of longer string
$m = y $.. length of shorter string
$L = LCS(x, y)$.. length of LCS
$ \Sigma $.. size of alphabet Σ
$\Delta = n - L$.. number of deletions in x
$\delta = m - L$.. number of deletions in y
M	.. number of <i>matching pairs</i>
d	.. number of <i>dominant pairs</i>

multivariate algorithms: $\tilde{O}(n + \min\{d, \delta m, \delta \Delta\})$

under SETH, this is **optimal** for any relations $m = \Theta(n^{\alpha_m}), L = \Theta(n^{\alpha_L}), \dots$



Grammar-Compressed Strings

String T given by **Straight-Line Program**:

[Abboud,Backurs,B.,Künnemann'17]

Nonterminals S_1, \dots, S_n

Each S_i is associated with a rule:

or $S_i \rightarrow c$ for some alphabet symbol/terminal c
 $S_i \rightarrow S_\ell S_r$ for some $\ell, r < i$

T is the string generated by S_n

compressed size n = number of nonterminals

uncompressed size N = length of T

$N \geq n$, and N can
be as large as $2^{\Omega(n)}$

Computing the LCS-length of strings x, y of
length N compressed to size n is in time $\tilde{O}(Nn)$...

... and not in $O((Nn)^{1-\varepsilon})$
time assuming SETH

[Tiskin'08] [Gawrychowski'12]

[Abboud,Backurs,B.,Künnemann'17]



Regular Expression Matching

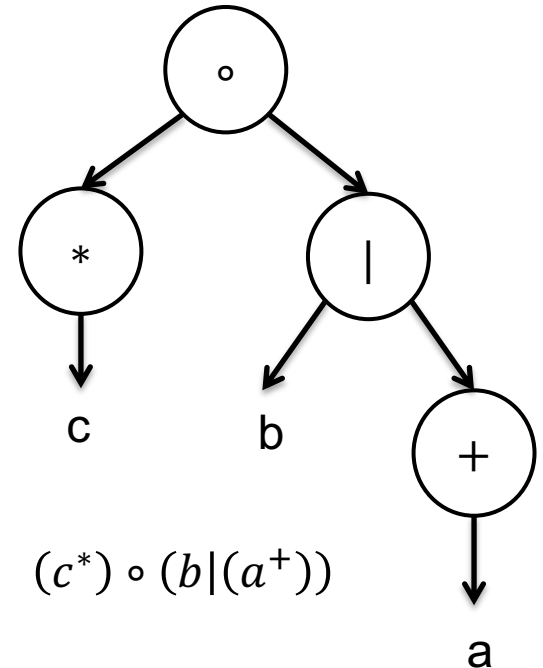
describe patterns using the operations:

- concatenation \circ
- alternation $|$ (OR)
- Kleene star $*$ (repetition with empty string)
- Kleene plus $+$ (repetition without empty string)

starting from alphabet symbols in Σ

great expressiveness: describe regular languages

standard tool to describe pattern matching problems on strings



Regular Expression Matching

$$n = |S|, m = |R|$$

Reg-Exp *Pattern Matching*:

given reg-exp R and string S , does some substring of S match R ?

$$O(nm)$$

[Thompson'68]

$$(nm)^{1-o(1)} \text{ under SETH}$$

[Backurs,Indyk'16]

Reg-Exp *Membership Testing*:

given reg-exp R and string S , does S match R ?

$$O(nm)$$

$$(nm)^{1-o(1)} \text{ under SETH}$$



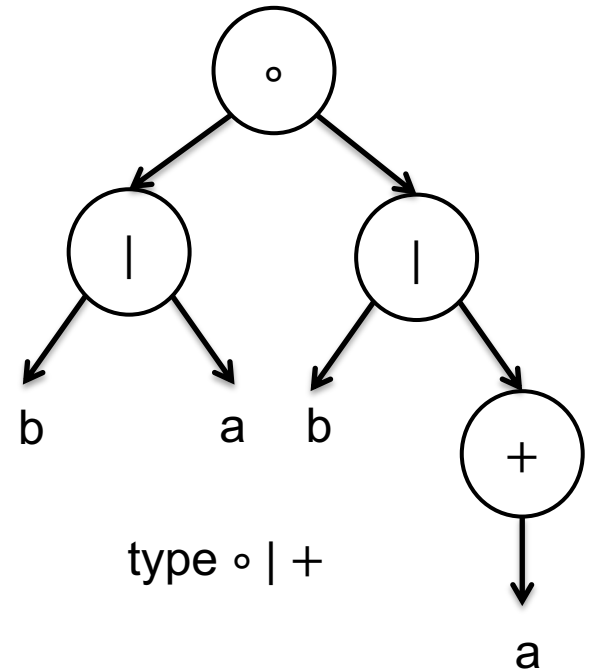
Homogeneous Regular Expressions

reg-exp is **homogeneous of type t** if:

- any internal node on level i has operation t_i
- the depth is at most $|t|$
- leaves can be at any level

restrict to homogeneous reg-exps of type t :

t -Pattern Matching, t -Membership Testing

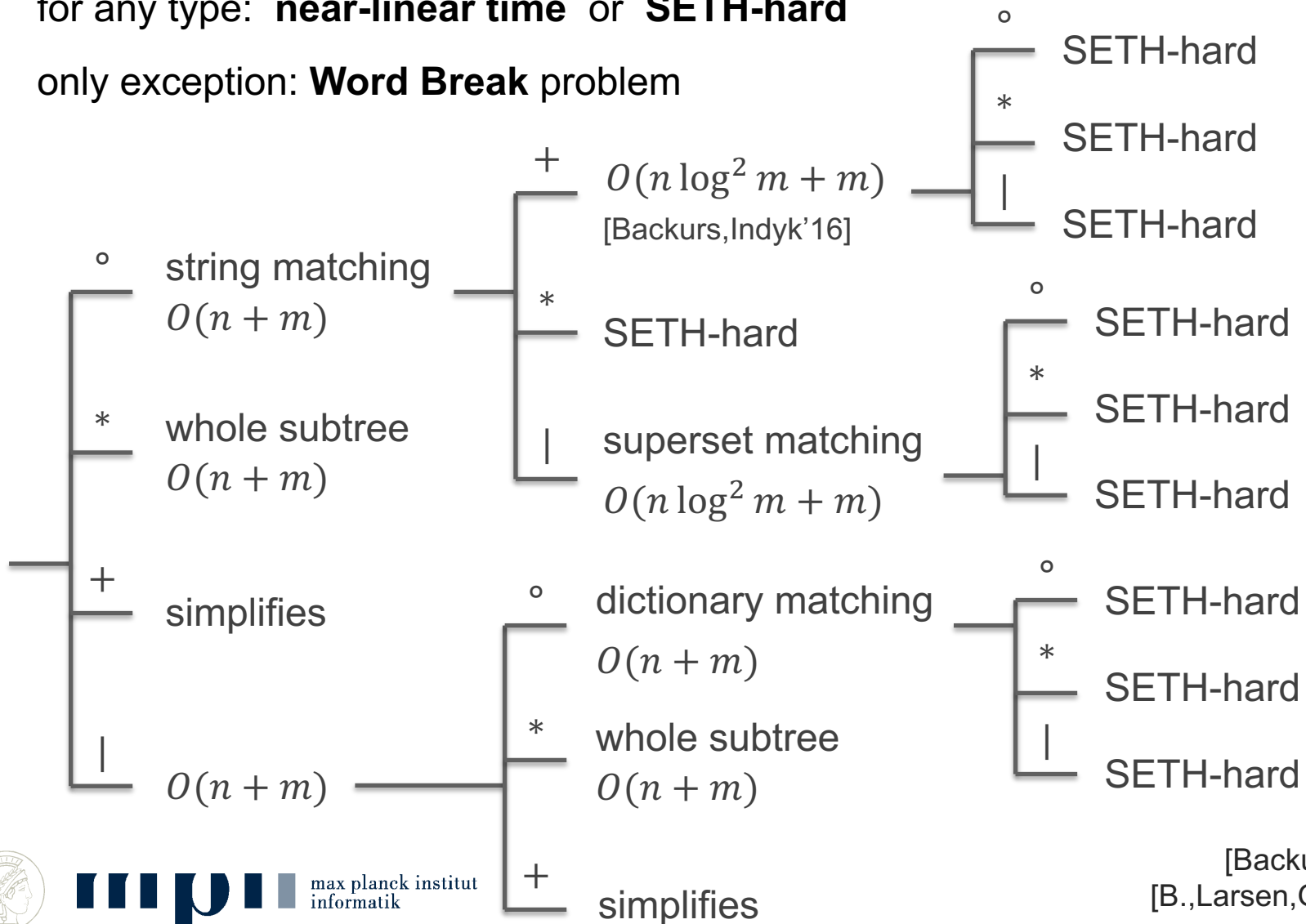


t -Pattern Matching and t -Membership Testing

complete characterization!

for any type: **near-linear time** or **SETH-hard**

only exception: **Word Break** problem



Word Break Problem

Word Break = +| ◦-Membership Testing

given string S and dictionary D

can S be split into words in D ?

$$n = |S|$$

$$m = \sum_{d \in D} |d|$$

$O(nm)$ trivial algorithm

$\tilde{O}(n\sqrt{m} + m)$ improved algorithm

$\tilde{O}(nm^{1/2-1/18} + m)$ [Backurs, Indyk'16]

$\tilde{O}(nm^{1/3} + m)$ [B., Larsen, Grønlund'17]

matching conditional lower bound from *k-Clique for combinatorial algorithms*

Word Break is the only **intermediate** t -Membership problem!



I. Easy Example

II. Longest Common Subsequence

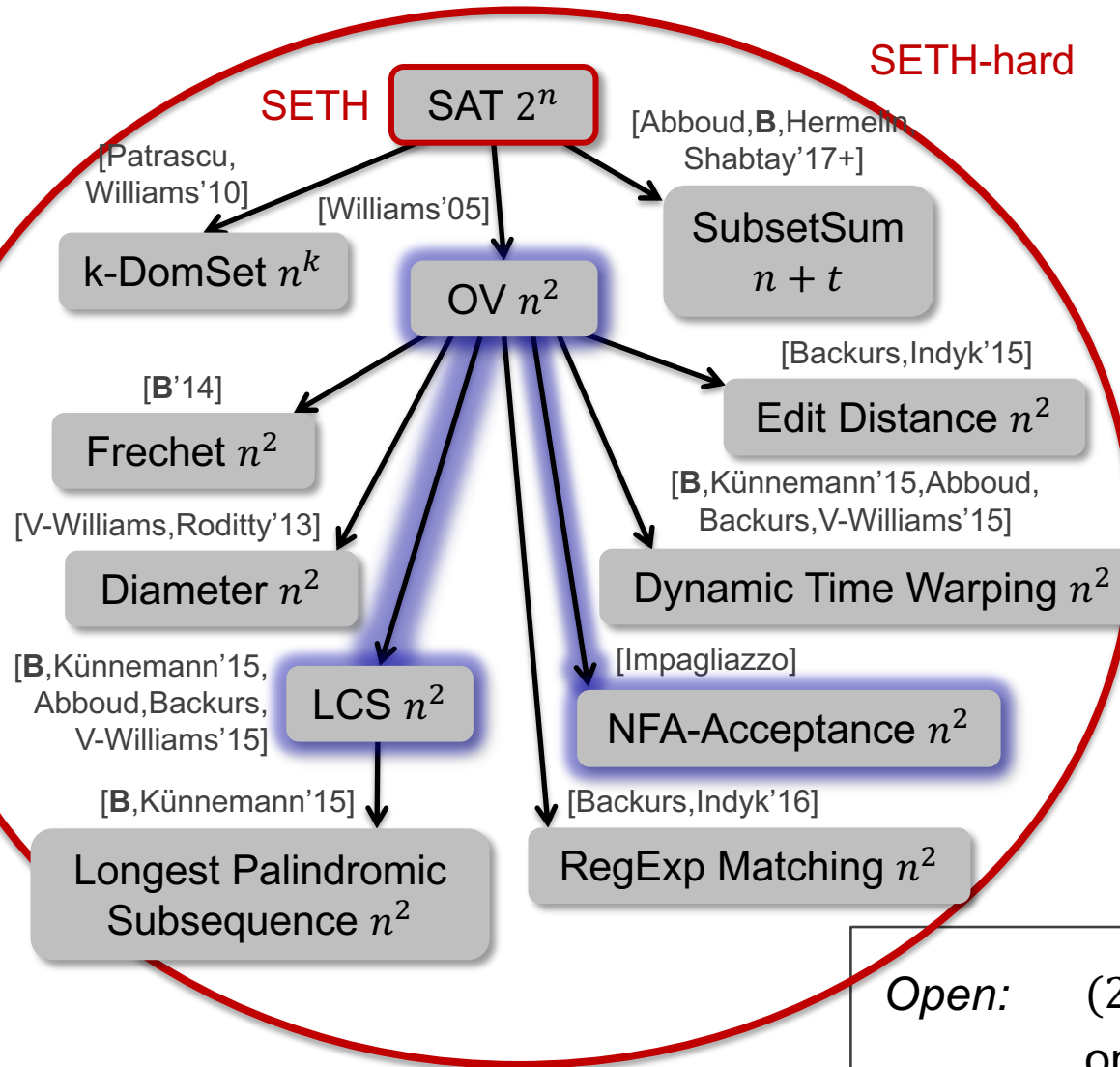
III. Hardness of Approximation

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Conclusion



we have seen:

- an involved lower bound
- coordinate gadgets,
- vector gadgets,
- normalized vector gadgets
- communication complexity (non-deterministic, randomized, one-sided)
- hardness of approximation in P

Open: $(2 - \epsilon)$ -approximation of **LCS** on binary strings in time $O(n^{2-\epsilon})$?

Hardness of approximation for non-product-like problems

