

Online Matrix-vector multiplication

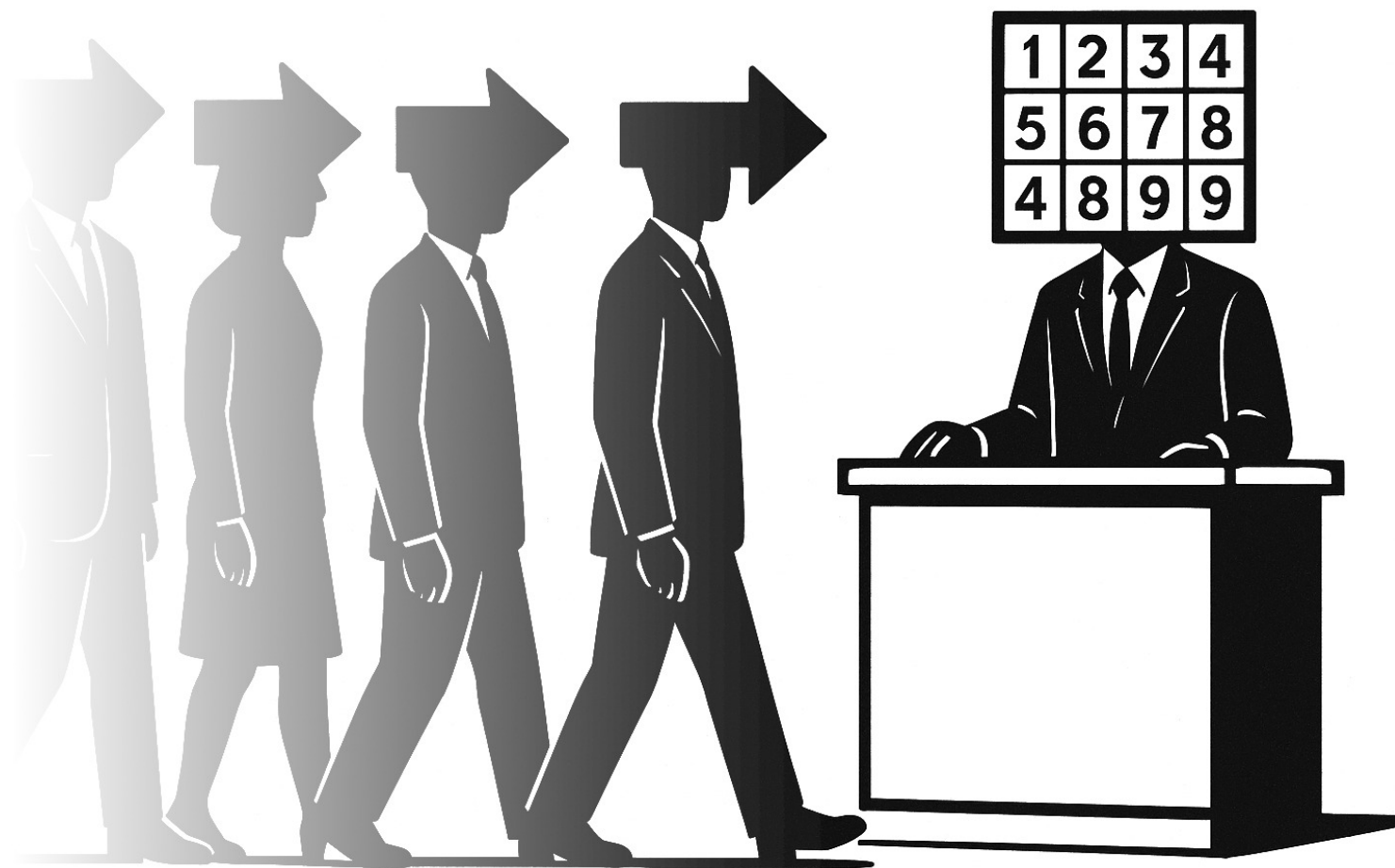
# Non-Boolean OMv

ONE MORE REASON TO BELIEVE LOWER BOUNDS FOR DYNAMIC PROBLEMS

Bingbing Hu

Adam Polak

UC San Diego Bocconi



# Dynamic algorithms

= data structures, but for **fancier** queries

Operations:

- **updates**

e.g., add edge, delete edge

- **queries**

e.g., find a path from source to  $v$ ,  
is the graph strongly connected?

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recompute  
from scratch

sublinear time  
update/query

polylog time  
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**fine-grained lower bounds**

# Fine-grained hypotheses

**3SUM**

**APSP**

**SETH**

**Zero- $\Delta$**

**k-Clique**

**OMv**

**k-SUM**

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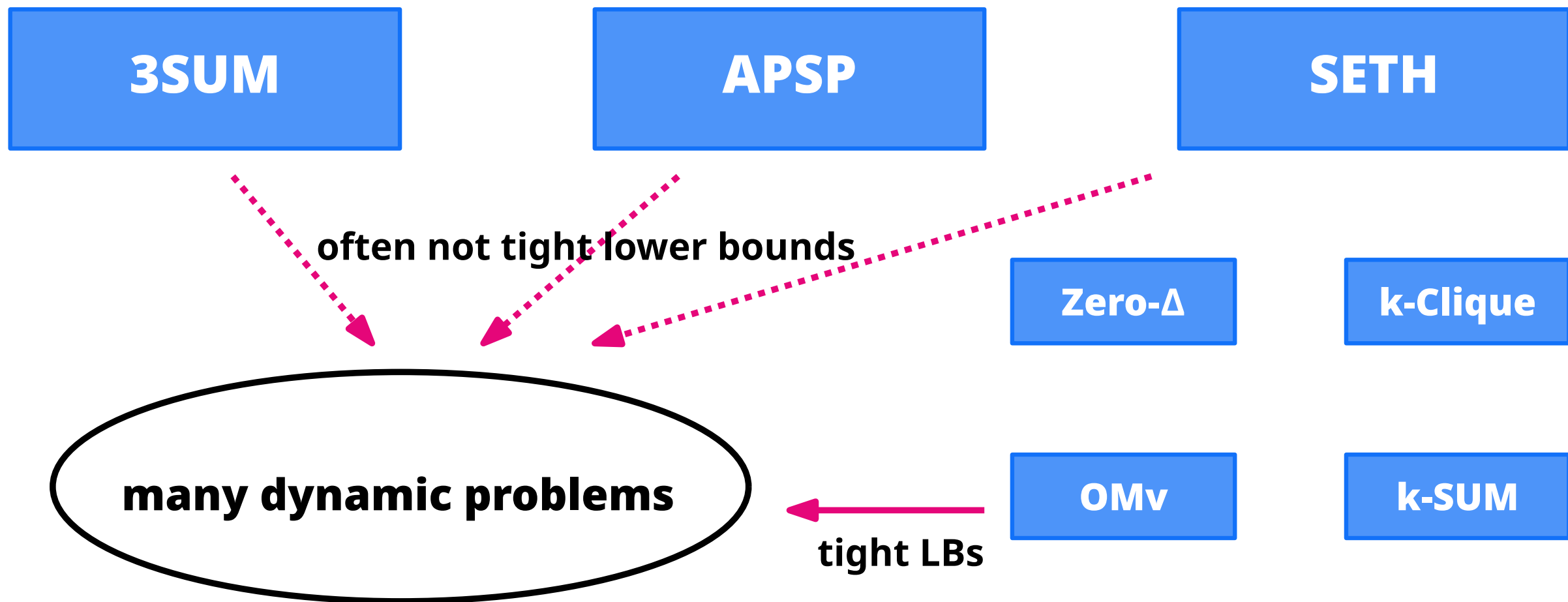
**many dynamic problems**

**OMv**

**k-SUM**

←  
tight LBs

# Fine-grained hypotheses



**Can we have *tight* lower bounds for dynamic problems based on a hypothesis that is *more believable than OMv*?**



# (Boolean) Online Matrix-vector multiplication (OMv)

[Henzinger–Krinninger–Nanongkai–Saranurak, STOC'15]

## Input:

Boolean  $n \times n$  matrix  $A$ ,

and  $n$  Boolean vectors  $v_1, v_2, \dots, v_n$  given **online**

## Output:

Boolean products  $Av_1, Av_2, \dots, Av_n$

must output  $Av_i$  **before being able to see**  $v_{i+1}$

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**OMv Hypothesis:** No  $O(n^{3-\epsilon})$  time algorithm for OMv

# Various matrix products (static)

**Easy**

$$O(n^{\omega}) < O(n^{2.372})$$

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- ▶ Boolean product

## Intermediate

$$O(n^{\frac{3+\omega}{2}}) < O(n^{2.686})$$

- ▶ Min-witness product
- ▶ Min-max product
- ▶ Dominance product
- ▶ Equality product
- ▶ Bounded monotone min-plus product

## Hard

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# Min-Max Online Matrix-vector multiplication

$$(M \circledast v)[i] := \min_j \max(M[i][j], v[j])$$

## Input:

Integer  $n \times n$  matrix  $A$ ,  
and  $n$  integer vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  given **online**

## Output:

**Min-Max** products  $M \circledast \mathbf{v}_1, M \circledast \mathbf{v}_2, \dots, M \circledast \mathbf{v}_n$   
must output  $M \circledast \mathbf{v}_i$  **before being able to see**  $\mathbf{v}_{i+1}$

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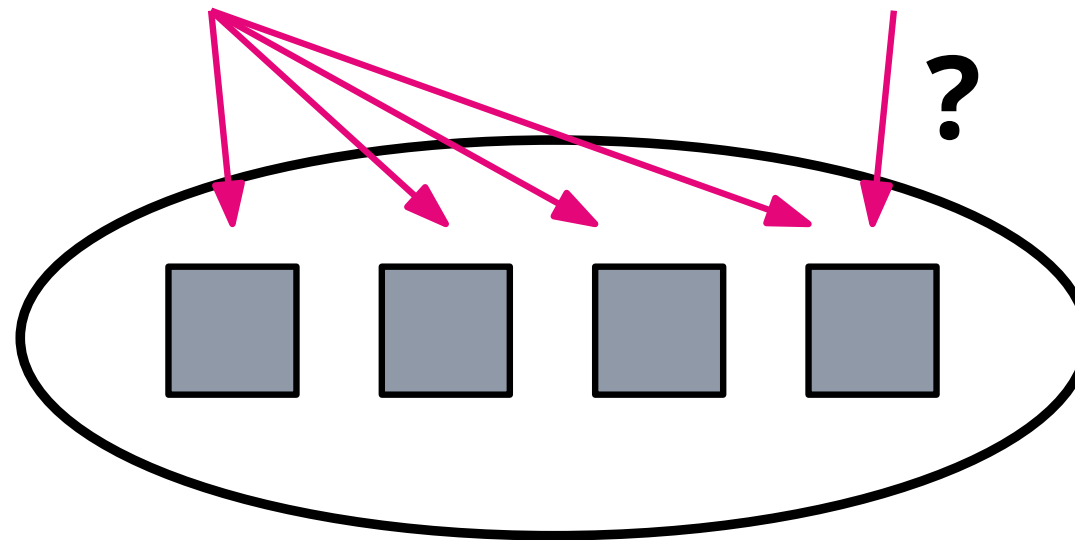
*a priori more believable* than **(Boolean-)OMv Hypothesis**



**Can we give **tight** reductions  
from **Min-Max-OMv** to those dynamic problems  
that have known tight reductions from (Boolean-)OMv?**

**Boolean-OMv**

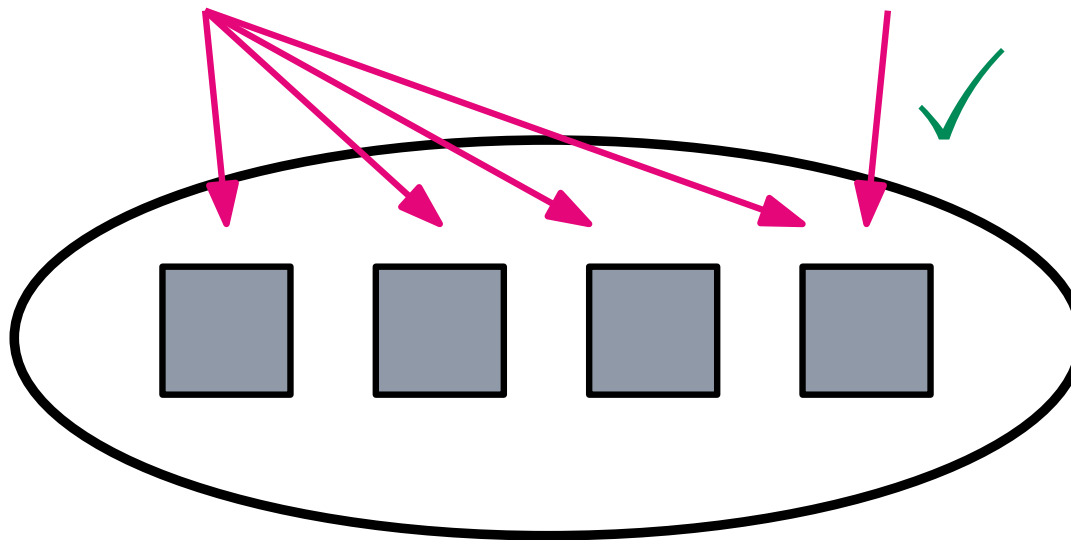
**Min-Max-OMv**



**many dynamic problems**

**Boolean-OMv**

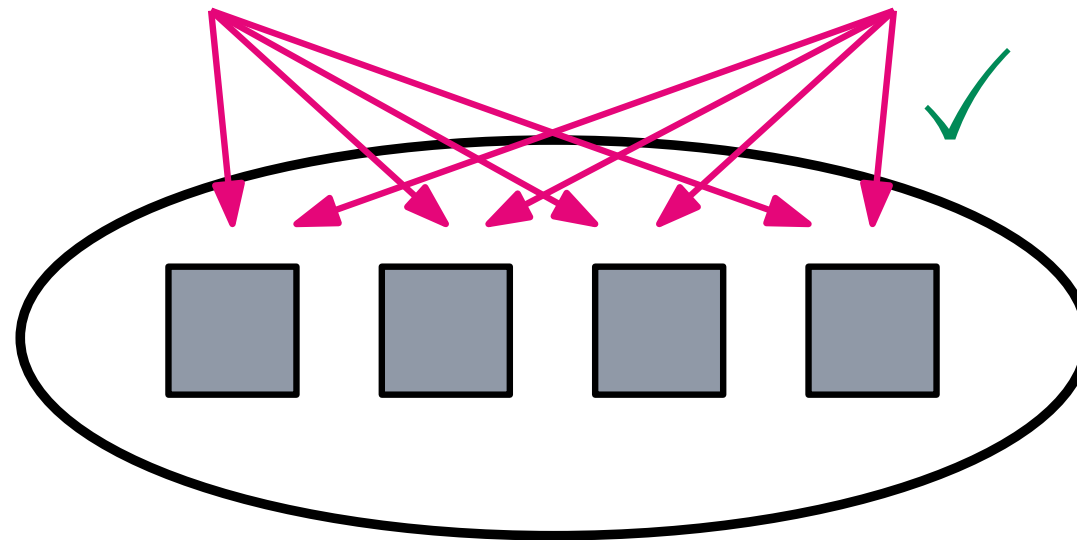
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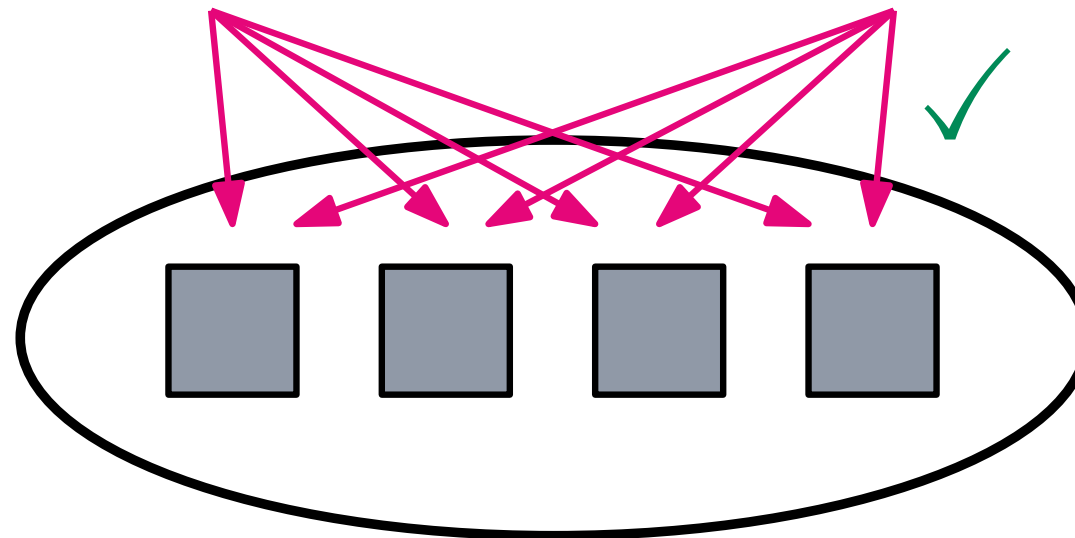
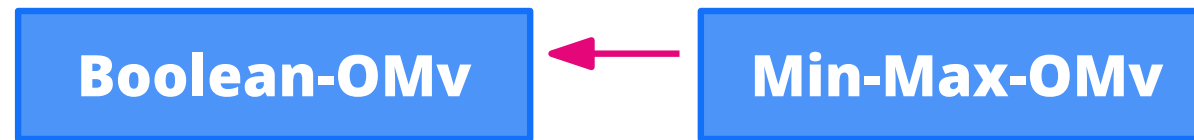
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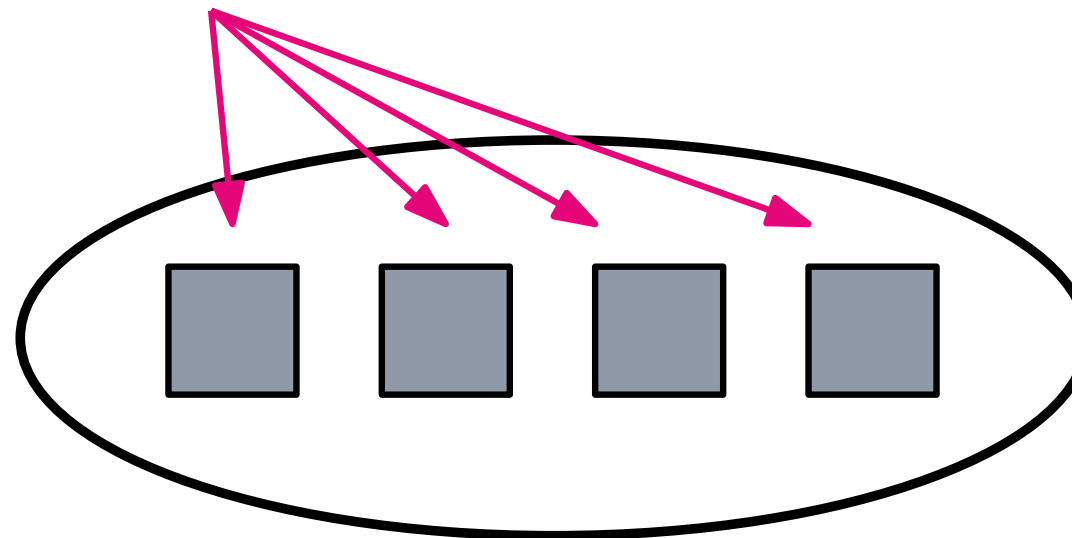
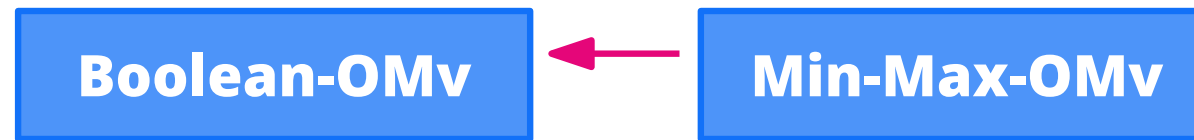
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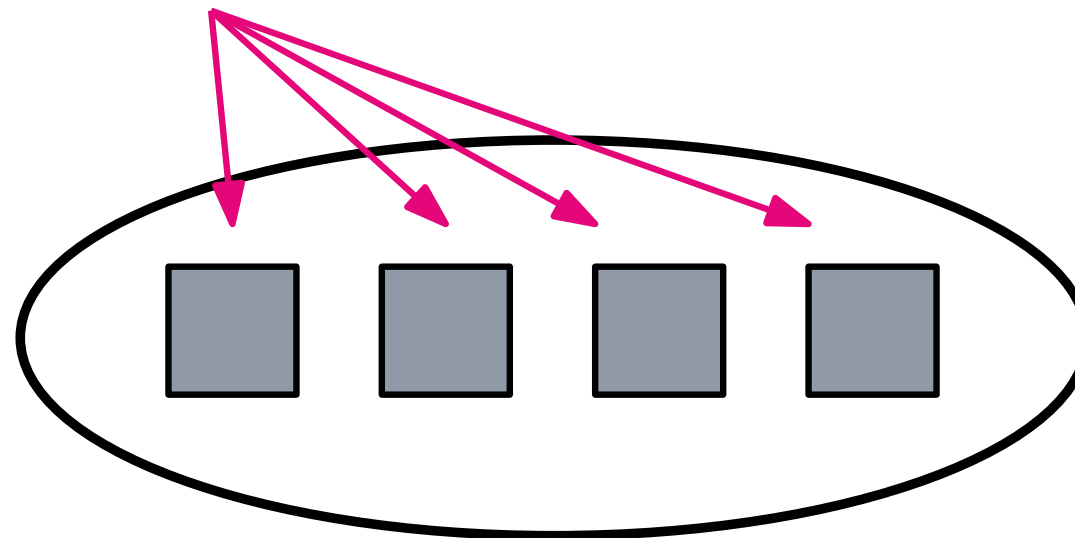
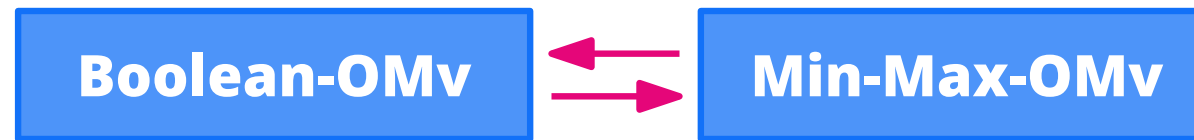
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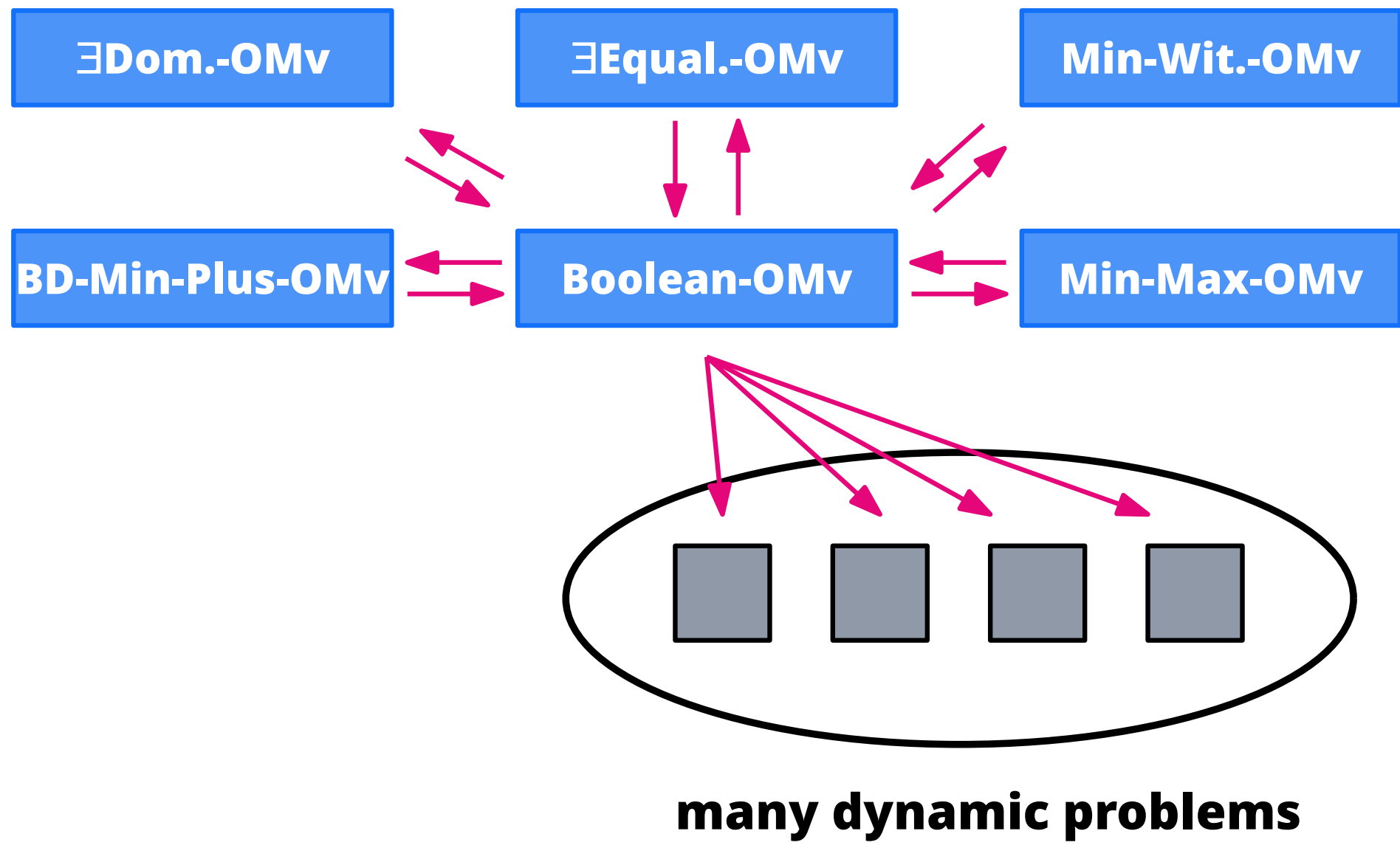
**many dynamic problems**



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**many dynamic problems**





# Our theorem

These problems **either all** have **truly subcubic** algorithms **or none** of them have:

- ▶ Boolean-OMv;  $(\exists_k M[i, k] \wedge v[k])$
- ▶  $\exists$ Equality-OMv;  $(\exists_k M[i, k] = v[k])$
- ▶  $\exists$ Dominance-OMv;  $(\exists_k M[i, k] \leq v[k])$
- ▶ Min-Witness-OMv;  $(\min \{k \mid M[i, k] \wedge v[k]\})$
- ▶ Min-Max-OMv;  $(\min_k \max\{M[i, k], v[k]\})$
- ▶ Bounded Monotone Min-Plus-OMv.  $(\min_k M[i, k] + v[k])$

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**Surprise?** Not in hindsight.

Known **static** algorithms  
in time  $O(n^{\frac{3+\omega}{2}})$

i.e.  $O(n^{f(\omega)})$  s.t.  
 $x < 3 \implies f(x) < 3$

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**Remark:** If (static) BMM has a subcubic *combinatorial* algorithm, then all these variants have such algorithms as well.

Open problem: Add a **counting** variant to the list.

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**THANK YOU!**