

Lecture 1: Definition and Complexity of Packing Problems

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at 10th Winter School on Computational Geometry
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Packing Problems

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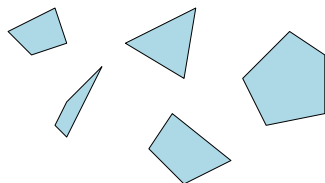
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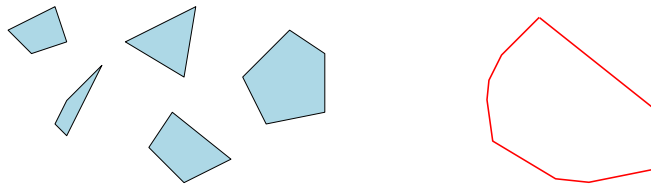
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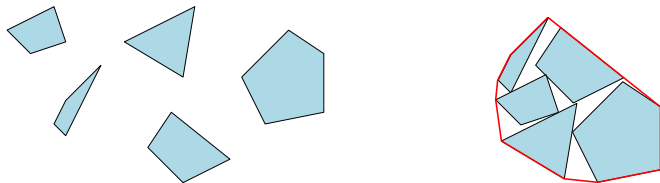
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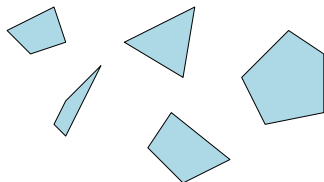
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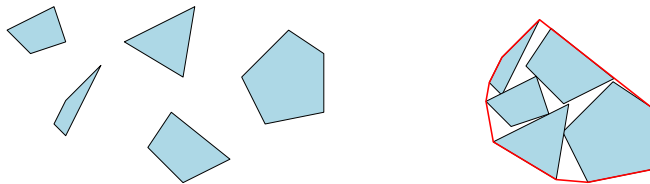
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- ▶ **objective function:**
area/volume, circumference/surface area, diameter

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It is well known: **$NP \subseteq PSPACE$** , the class of all problems solvable in polynomial *space*.

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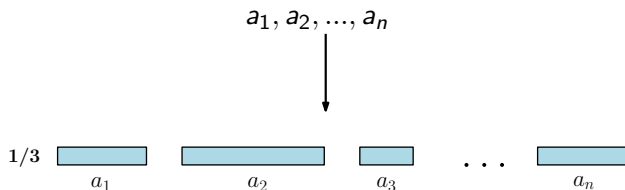
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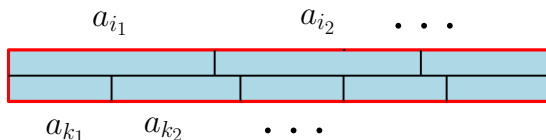
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partition $a_{i_1} + \dots + a_{i_l} = a_{k_1} + \dots + a_{k_m}$ is possible, iff



perfect packing is possible

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- ▶ Claim: PARTITION is efficiently reducible to PAR.
- ▶ Consequently: PAR is NP-complete.
- ▶ also, it follows or can easily be shown:
 - ▶ packing of rectangles/arbitrary polygons into rectangular/polygonal containers under translation/rigid motion is NP-hard.
 - ▶ for rectangular/convex polygonal containers the corresponding optimization problems are NP-hard.

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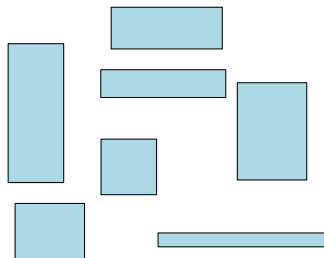
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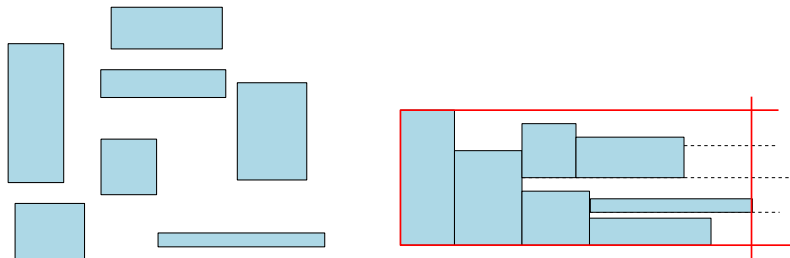
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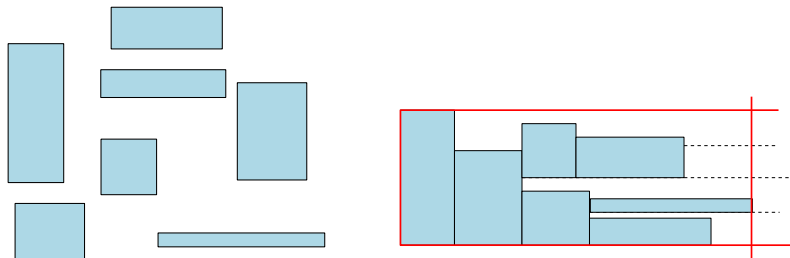
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also possible:

3-approximation for packing axes-parallel rectangles under **rigid motion**

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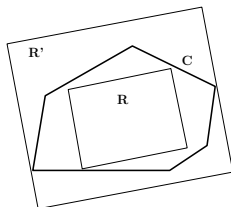
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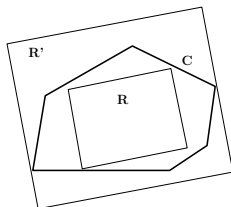
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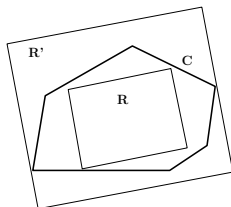


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It follows: Any α -approximative packing algorithm for rectangular objects can be used to get a 4α -approximation algorithm for convex polygonal objects.