

# The Rectilinear Marco Polo Problem

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<sup>1</sup>University of California, Irvine

<sup>2</sup>University of Rochester

CCCG, 2025



# The Marco Polo Problem I

- Point of Interest (POI)  $x$
- $x$  within distance  $n$  from origin

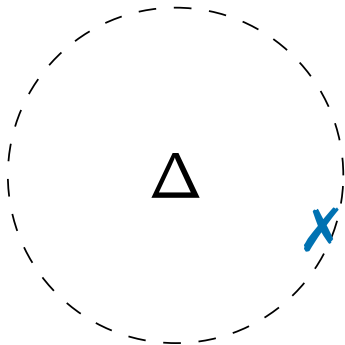


Figure 1: A search area in  $L_2$ .

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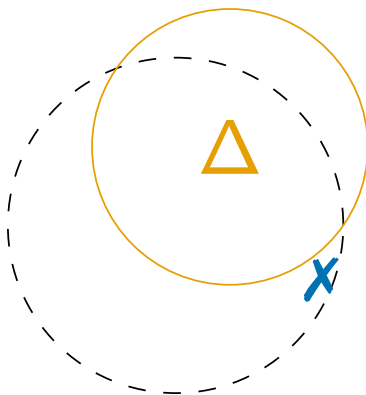


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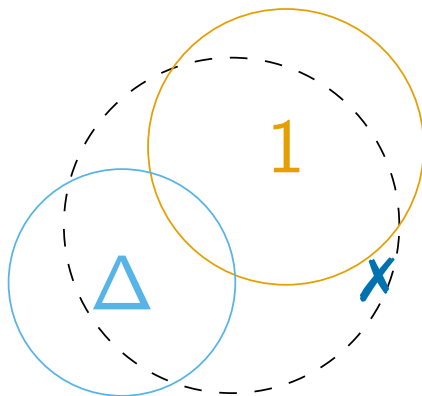


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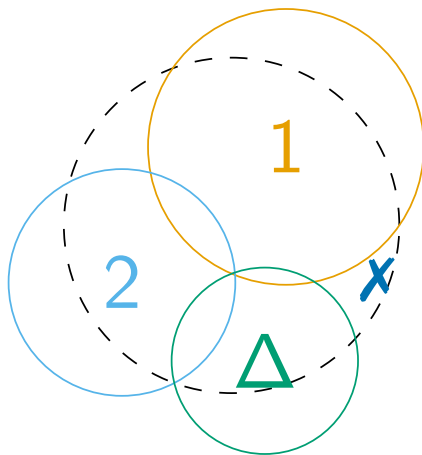


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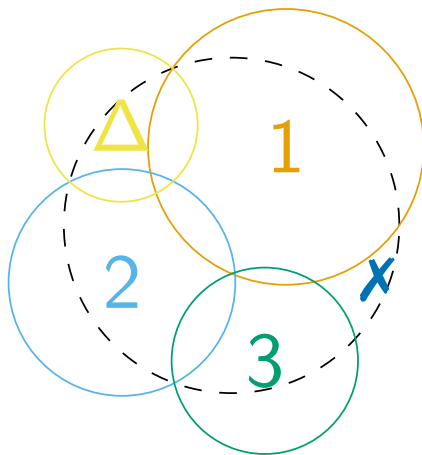


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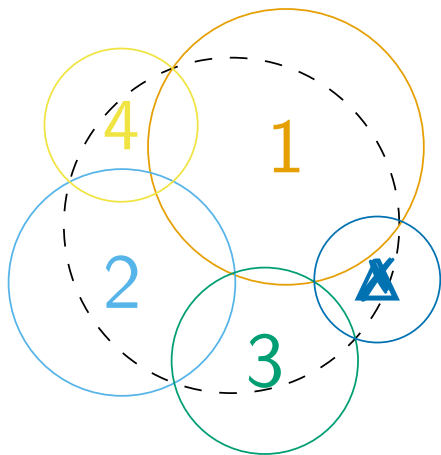


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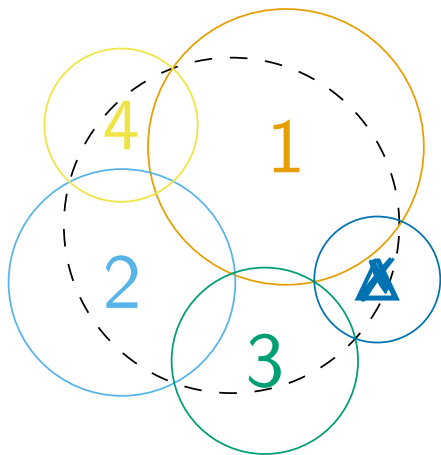


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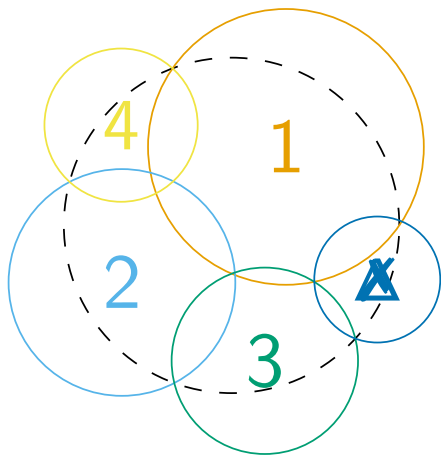


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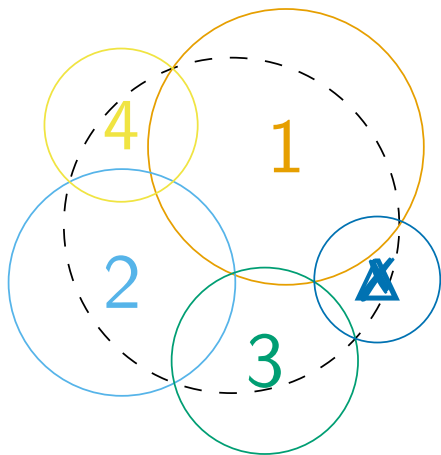


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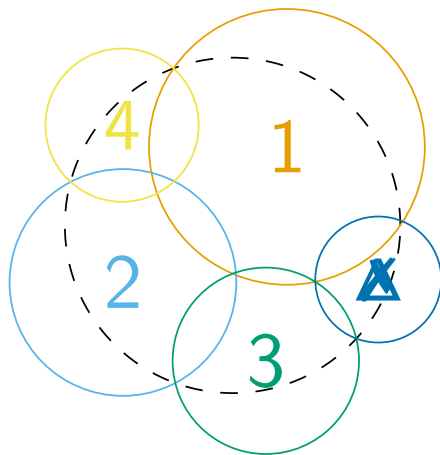


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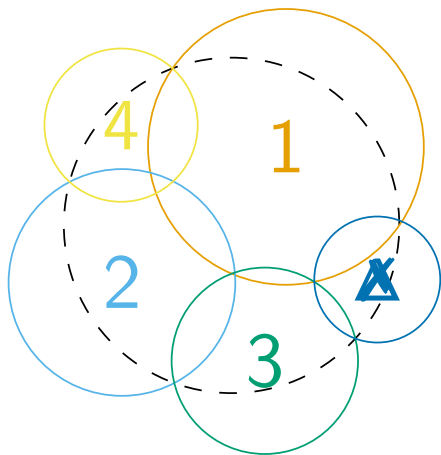


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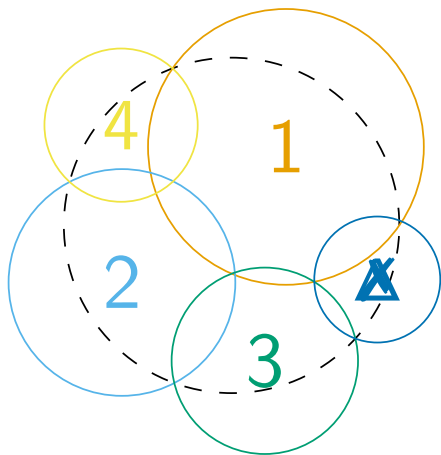


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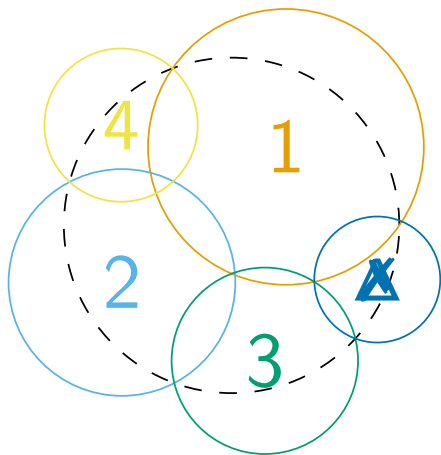


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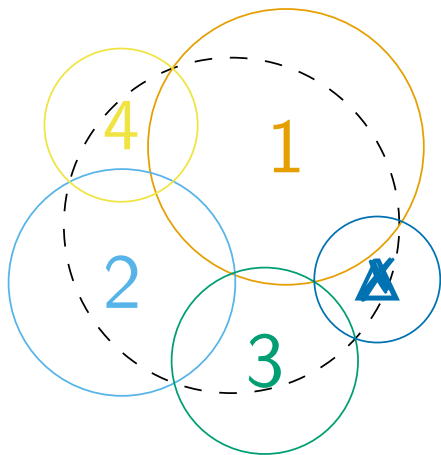


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- Variants:

- ☒ # of POIs present ( $k$ )
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- ☐ Probe response (T/F,  $d, i$ )
- ☐  $\Delta$ 's memory if any
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- Effectiveness metrics:

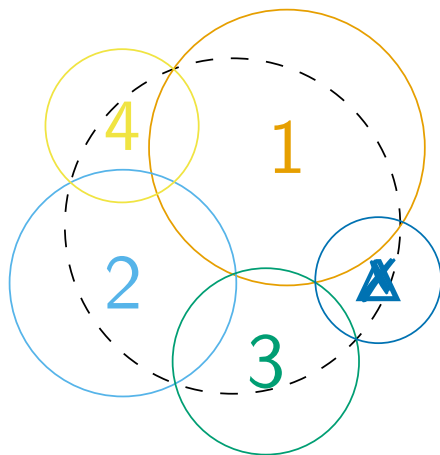


Figure 2: A search area in  $L_2$ .



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- ☐ # of probes,  $P(n)$

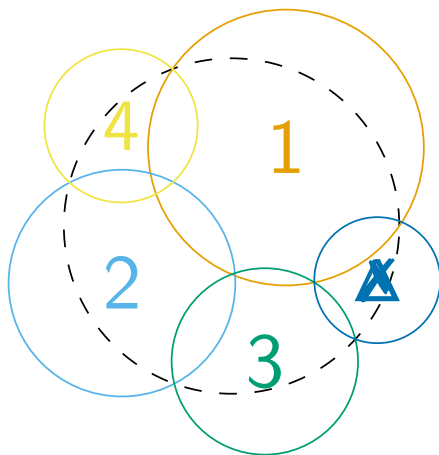


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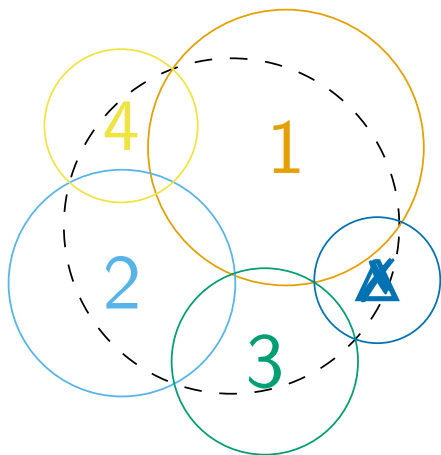


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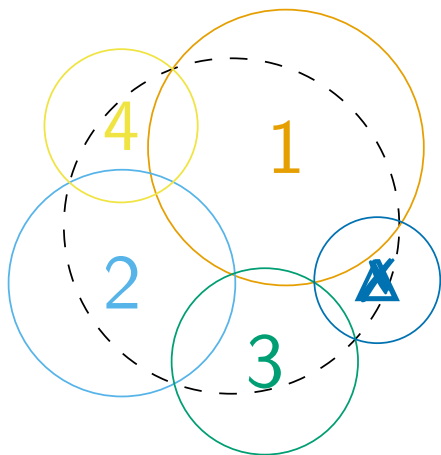


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- ☒ # of POI responses,  $R(n)$ 
  - Input sensitivity?

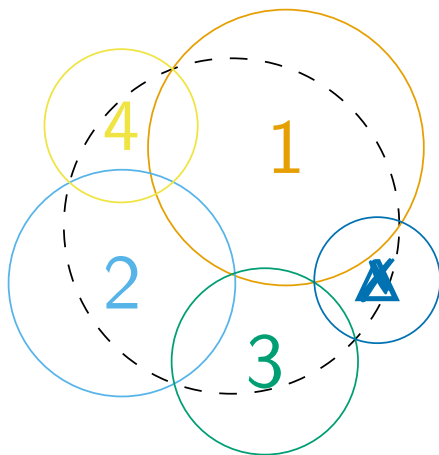


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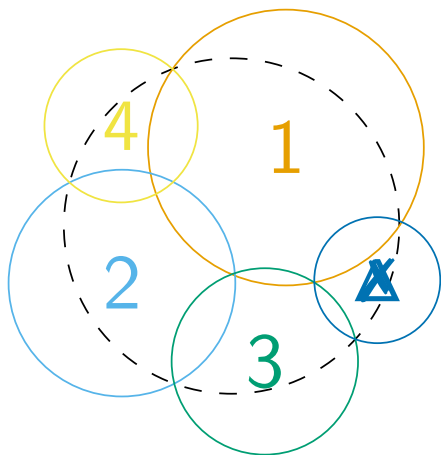


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- Input sensitivity?

- ☒ TSP tour length,  $\text{OPT}$
- ☐ Dist. to nearest POI,  $\delta_{\min}$

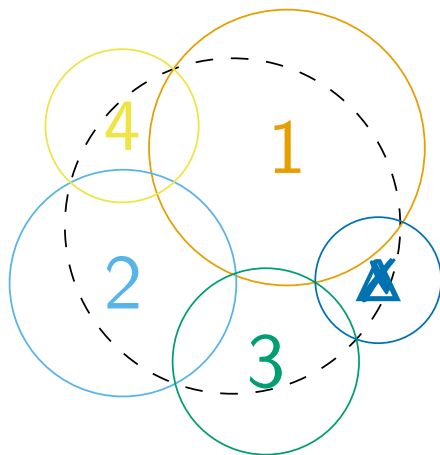


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- Input sensitivity?

- ☒ TSP tour length, OPT
- ☐ Dist. to nearest POI,  $\delta_{\min}$

- ☐ Simplicity / practicality

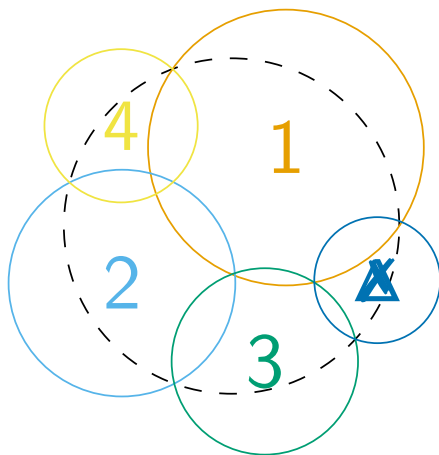


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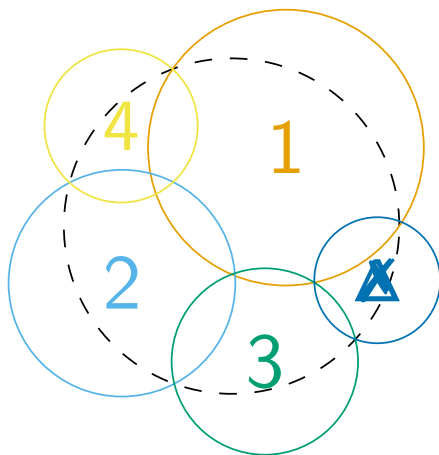
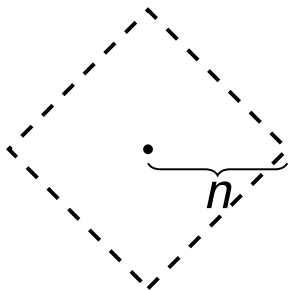


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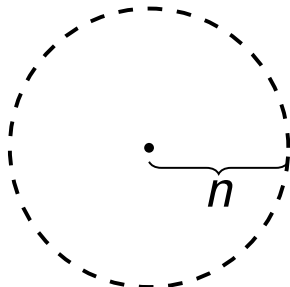
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- 1D: All identical



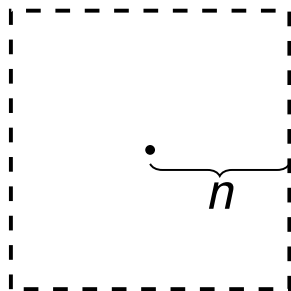
$$\|x\|_1 = \sum_{i=1}^n |x_i|$$

Figure 3:  $L_1$  search area.



$$\|x\|_2 = \sqrt{\sum_{i=1}^n x_i^2}$$

Figure 4:  $L_2$  search area.

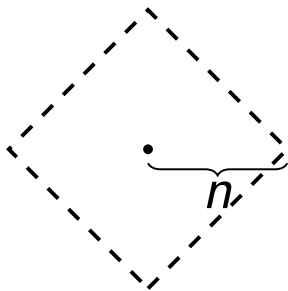


$$\|x\|_\infty = \max_i |x_i|$$

Figure 5:  $L_\infty$  search area.

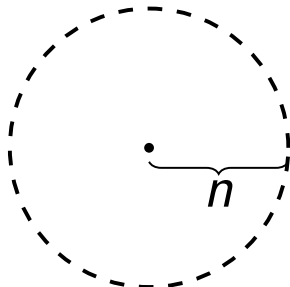
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- 1D: All identical
- 2D:  $L_1$  and  $L_\infty$  geometrically similar



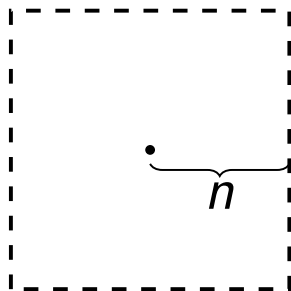
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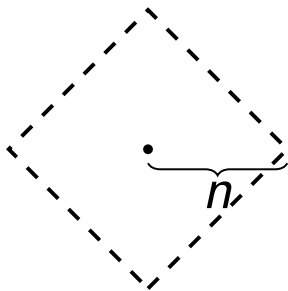


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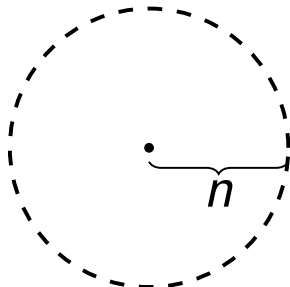
# Rectilinear Distances

- 1D: All identical
- 2D:  $L_1$  and  $L_\infty$  geometrically similar
- $\geq 3D$ : All different



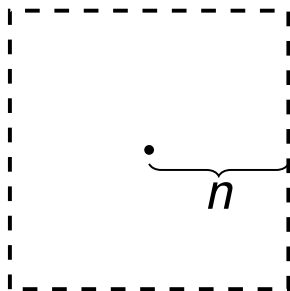
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# Rectilinear Strategies: Quadrant Algorithm

- Consider a  $2 \times 2$  lattice

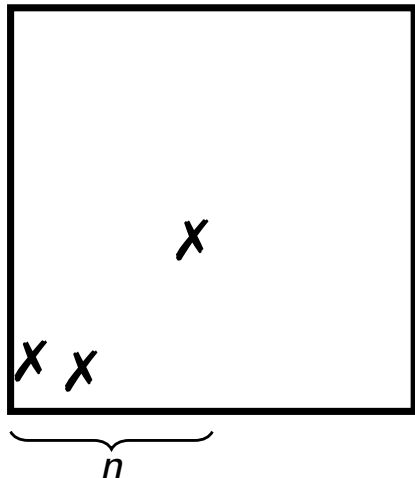


Figure 6: Quadrant Search

# Rectilinear Strategies: Quadrant Algorithm

- Consider a  $2 \times 2$  lattice
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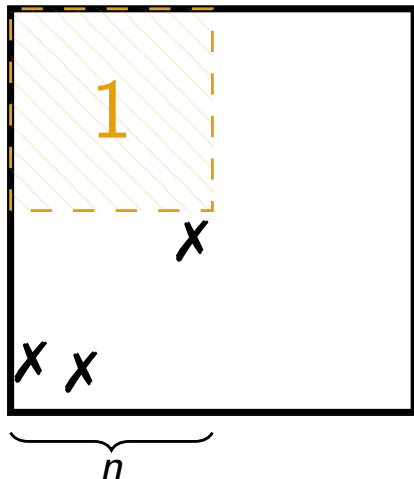


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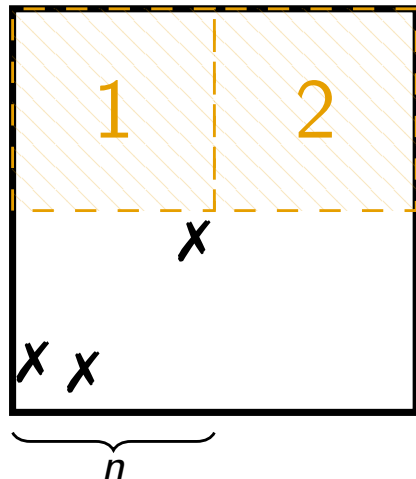


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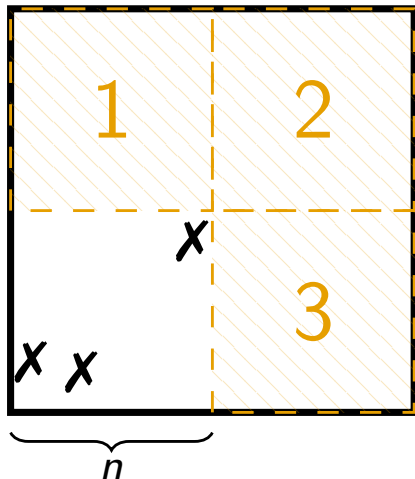


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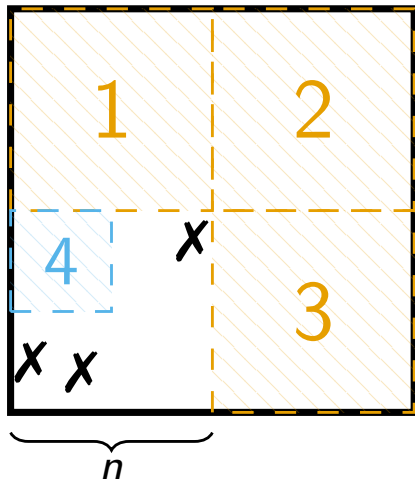


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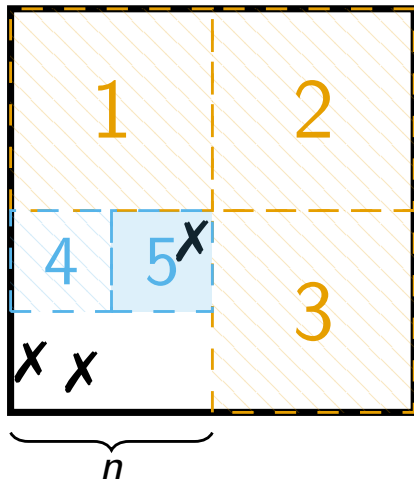


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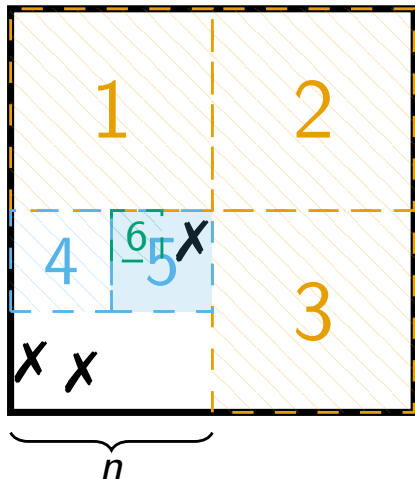


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- Continue recursively!

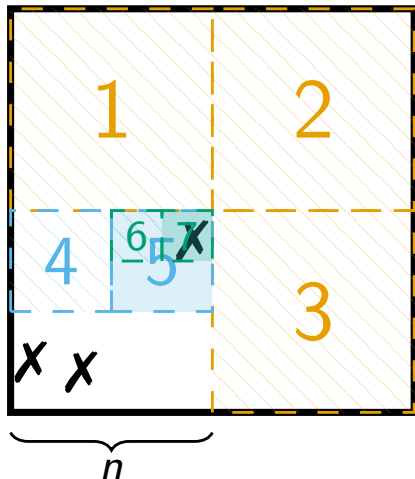


Figure 6: Quadrant Search

# Rectilinear Strategies: Quadrant Algorithm

- Consider a  $2 \times 2$  lattice
- Probe each quadrant...
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- Continue recursively!
- After 3 probes... distance halved!

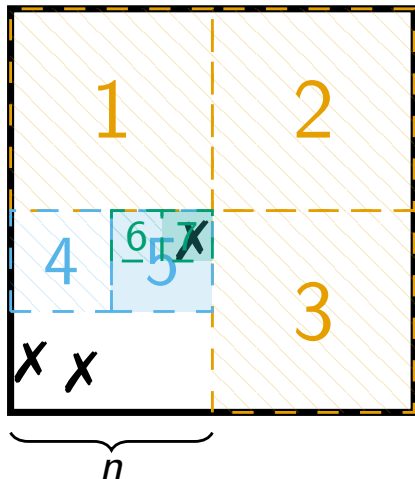


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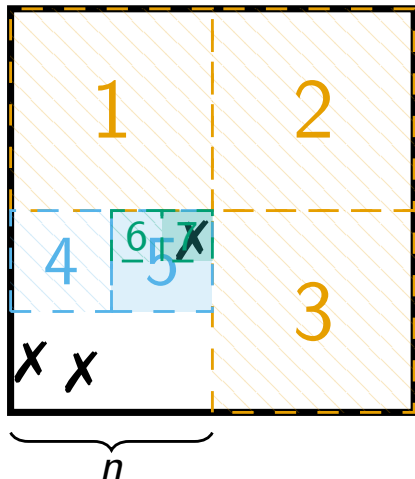


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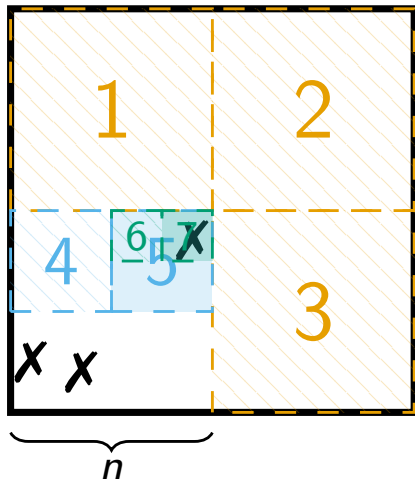


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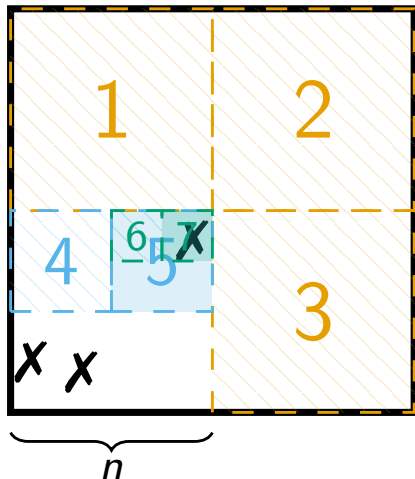


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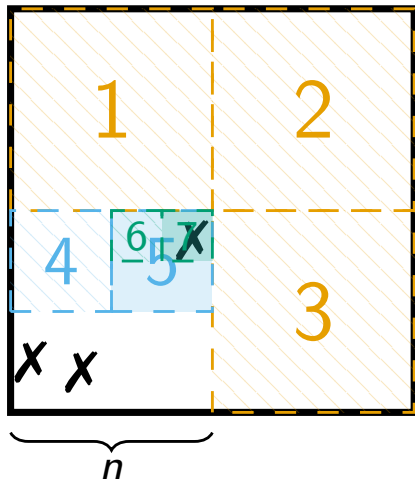


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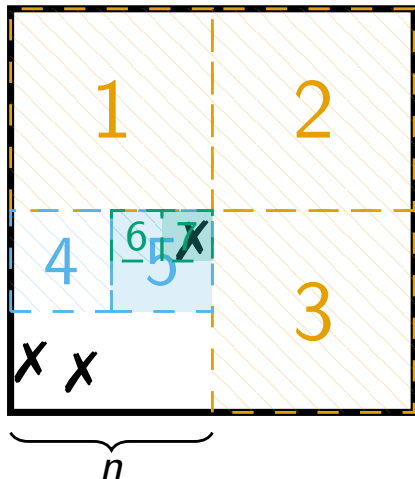


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- Distance traveled?
- $D(n) \leq 6n$

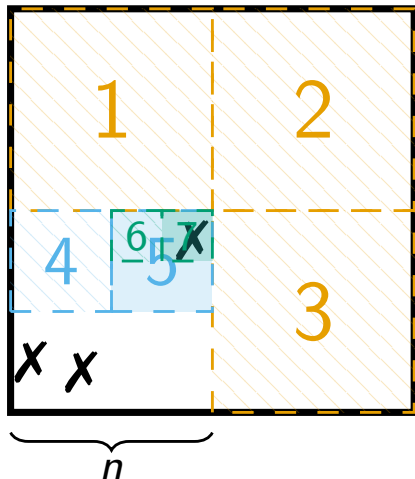


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## 2D Domino Algorithm

- Quadrant:  $P(n) \leq 3 \lceil \log n \rceil$

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# 2D Domino Algorithm

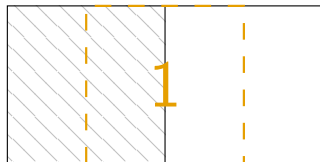
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## 1 Probe center

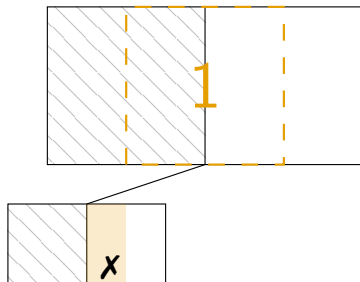




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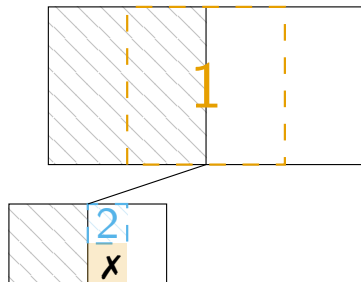
- 1 Probe center
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  - a If 1st probe succeeded...



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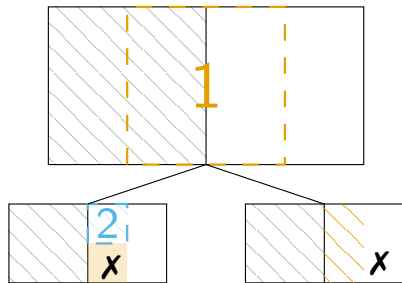
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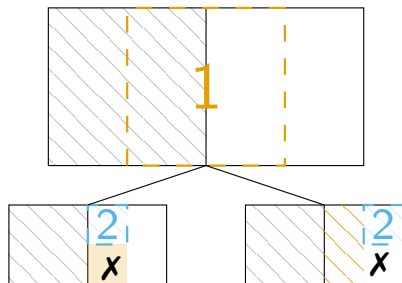
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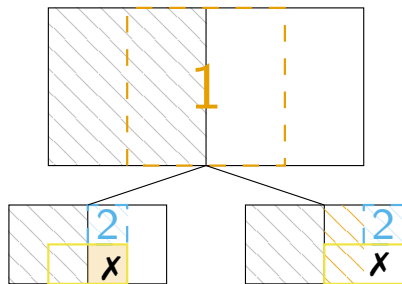
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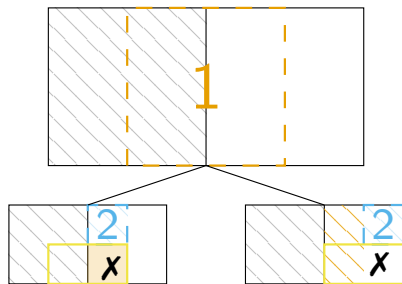
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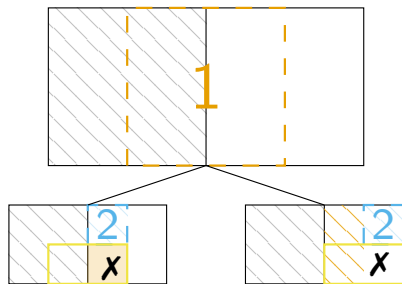
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# 2D Domino Algorithm

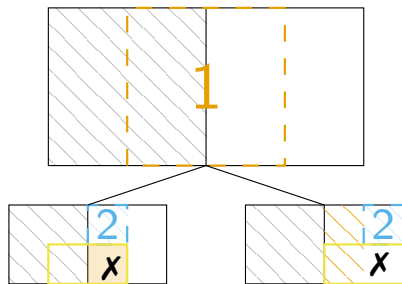
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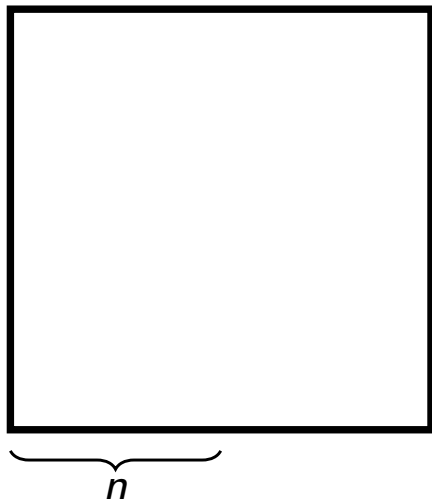
- Either way...  $n \times n/2$  ‘domino’
- Two probes  $\rightarrow$  halve dimensions
- $P(n) = 2 \lceil \log n \rceil$
- How to reach domino?





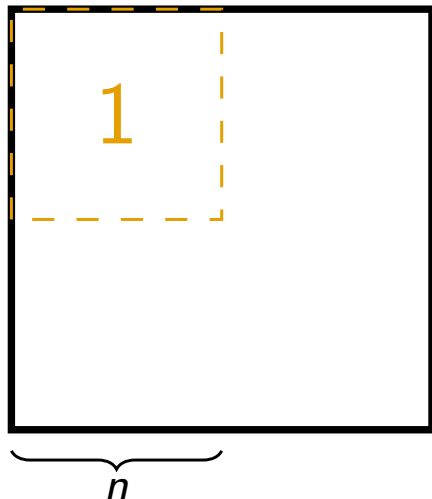
# Rectilinear Strategies: 2D Domino Algorithm – cont'd

- How to reach domino?



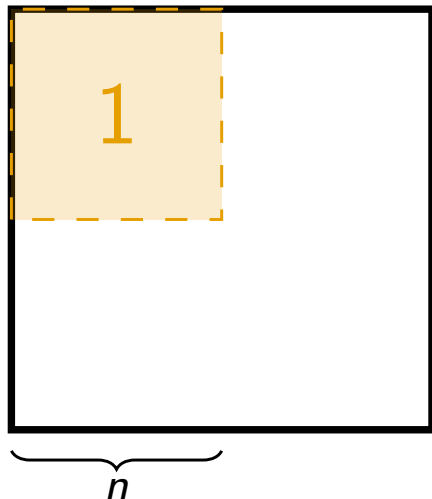
# Rectilinear Strategies: 2D Domino Algorithm – cont'd

- How to reach domino?
- Start quadrant search...



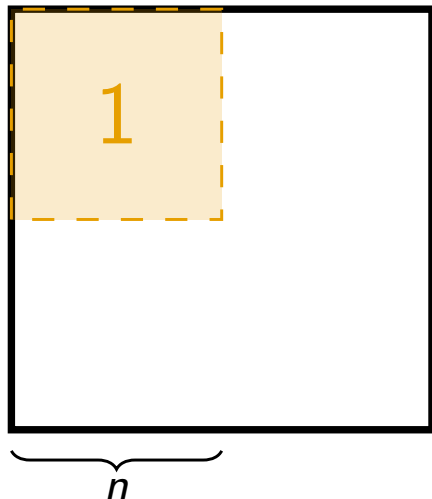
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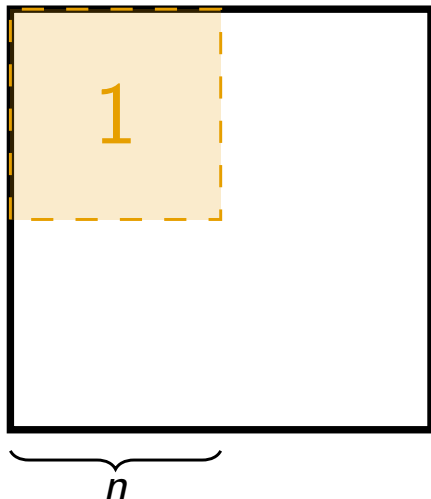
# Rectilinear Strategies: 2D Domino Algorithm – cont'd

- How to reach domino?
- Start quadrant search...
- If first probe succeeds...
- Reduce area by factor of 4!



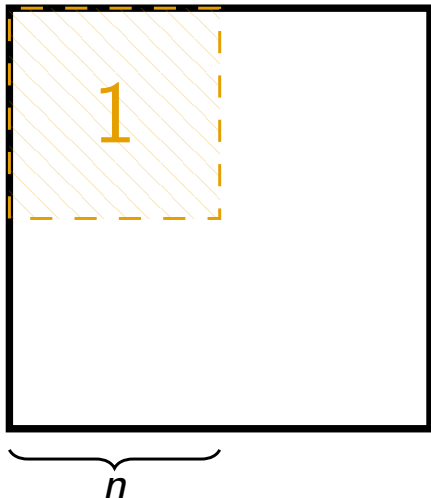
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- How to reach domino?
- Start quadrant search...
- If first probe succeeds...
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- Just recurse and 😊



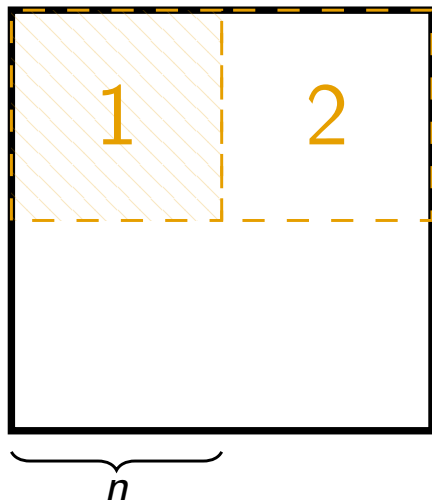
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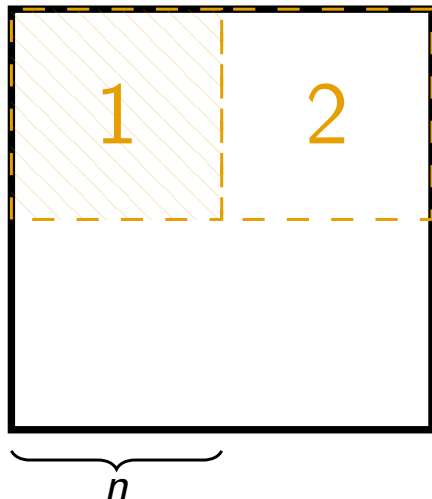
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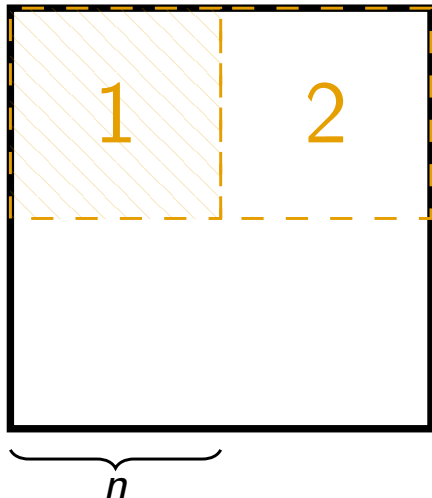
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- $P(n) \leq 2^{\lceil \log n \rceil} + 1$





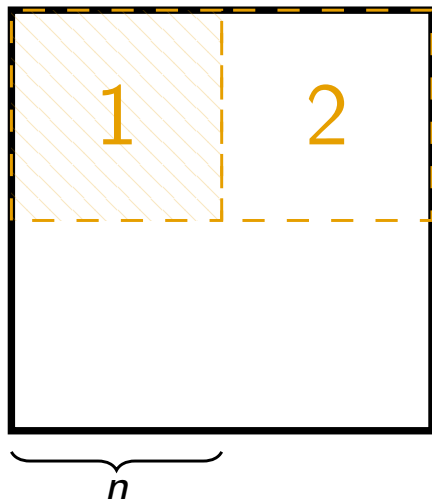
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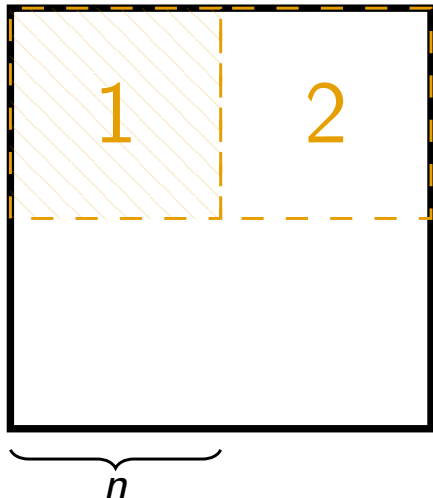
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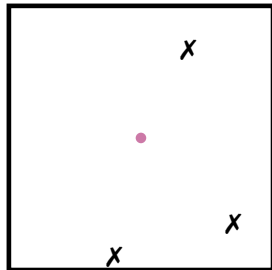
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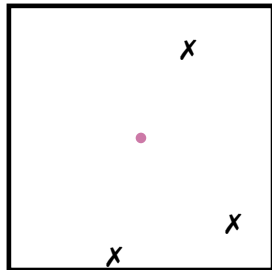
# Rectilinear Strategies: Central Binary Search (CBS)

- Recall:  $D(n) \in \mathcal{O}(n)$



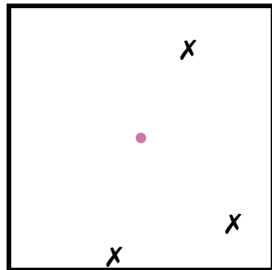
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- Recall:  $D(n) \in \mathcal{O}(n)$
- If POI is nearby?



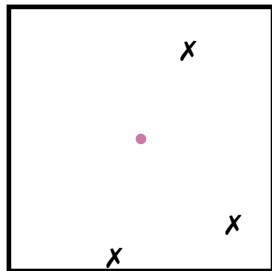
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- Recall:  $D(n) \in \mathcal{O}(n)$
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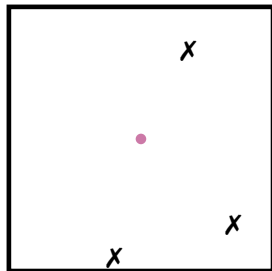
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# Rectilinear Strategies: Central Binary Search (CBS)

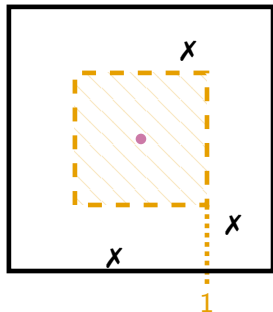
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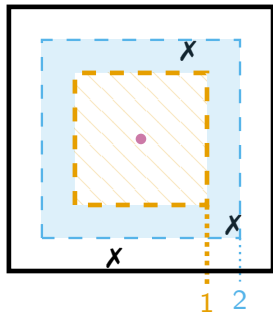
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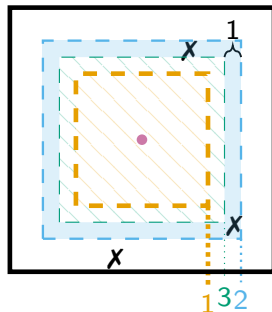
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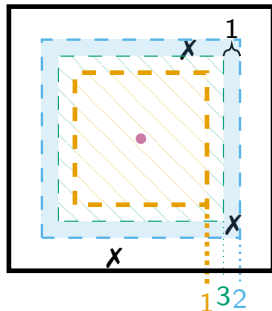
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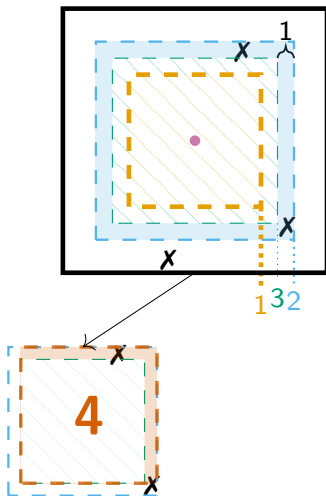
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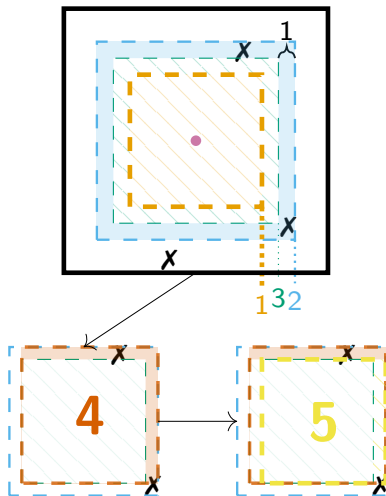
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- $\lceil \log n \rceil$  probes, 0 distance
- Determine edge containing POI.



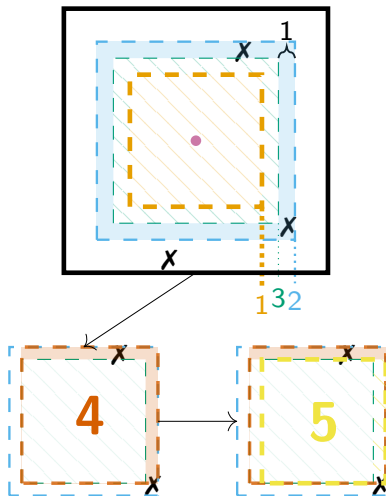
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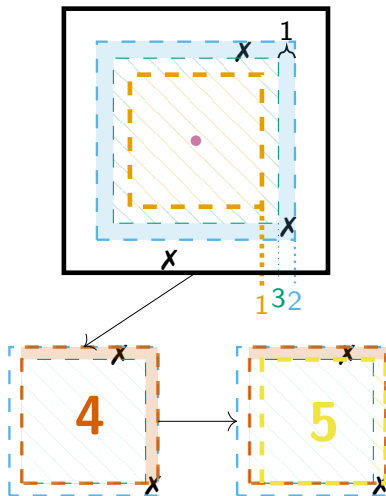
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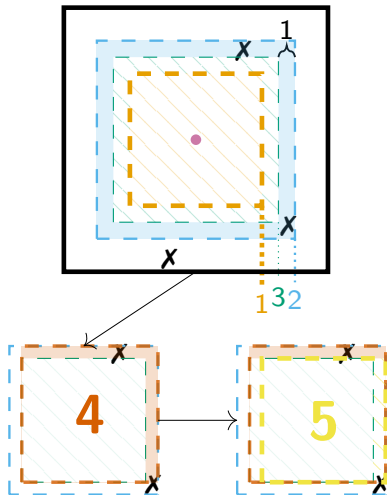
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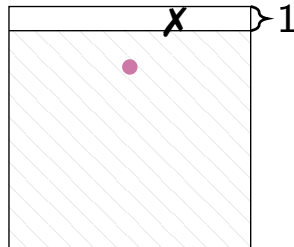
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- Recall:  $D(n) \in \mathcal{O}(n)$
- If POI is nearby? **still  $\mathcal{O}(n)$**
- Nearest POI at distance  $\delta_{\min}$
- Can we do  $\mathcal{O}(\delta_{\min})$ ?
- Perform binary search...
- Stop when width-1 shell\*
- $\lceil \log n \rceil$  probes, 0 distance
- Determine edge containing POI..
- 2 probes – near origin!
- $\mathcal{O}(1)$  distance
- What now?



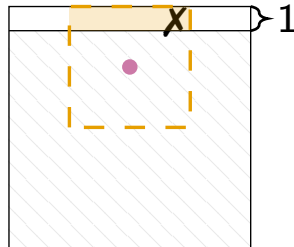
# Rectilinear Strategies: CBS – 2nd Phase

- What now?



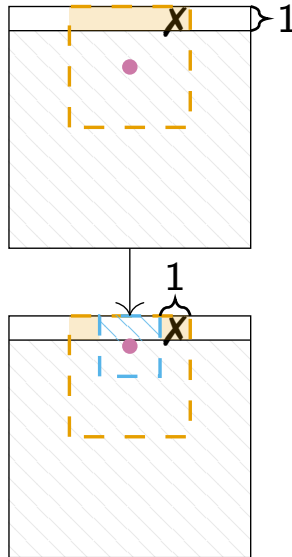
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- What now?
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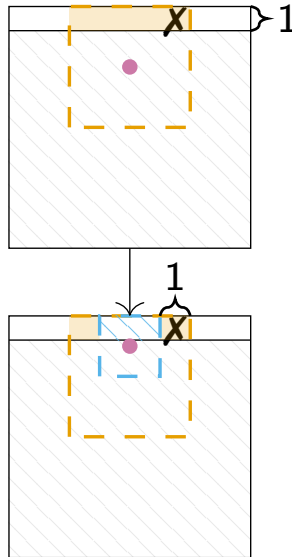
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- What now?
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- At most  $\lceil \log n \rceil$  probes



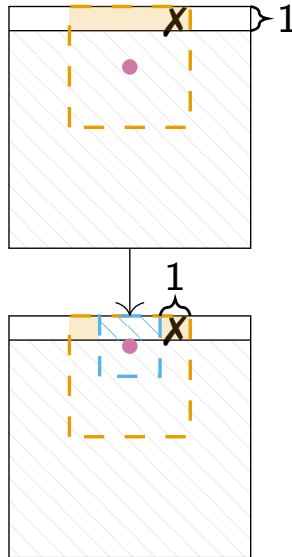
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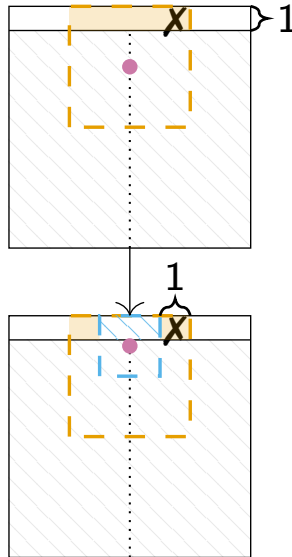
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- What now?
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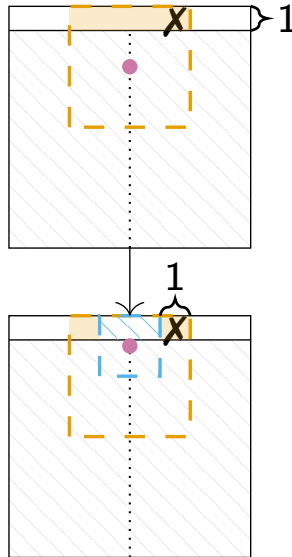
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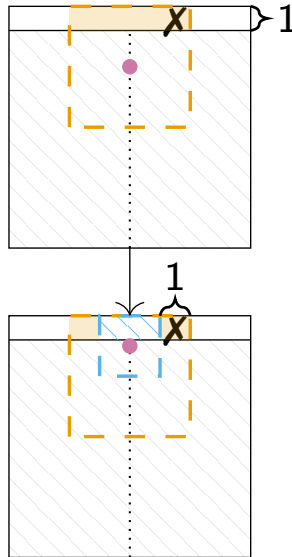
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- First probe,  $\leq \delta_{\min}/2$





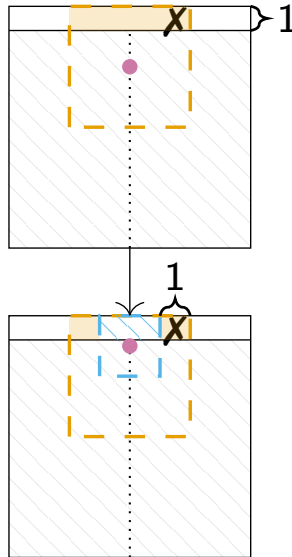
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- At most  $\lceil \log n \rceil$  probes
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- Second,  $\leq \delta_{\min}/4$



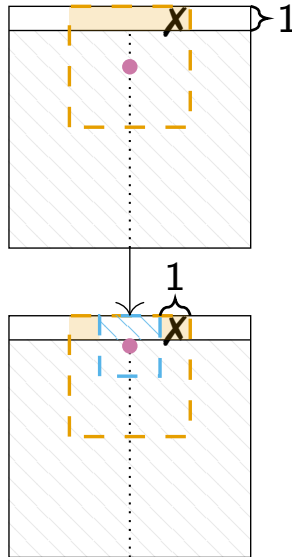
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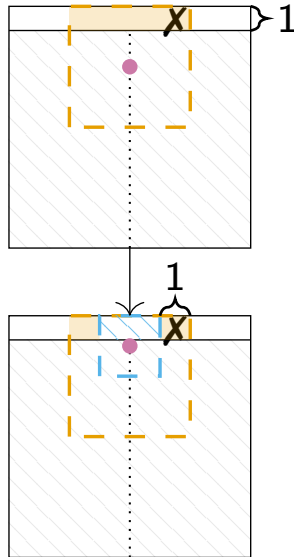
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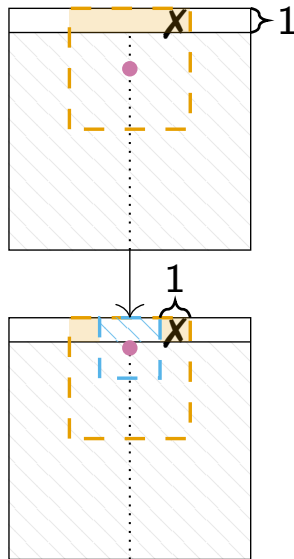
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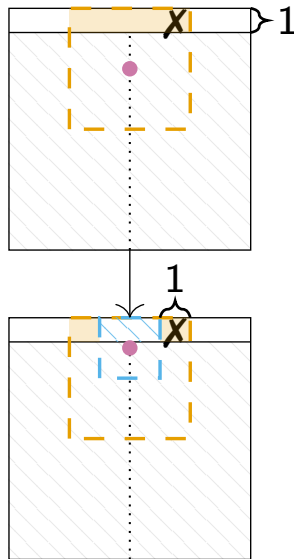
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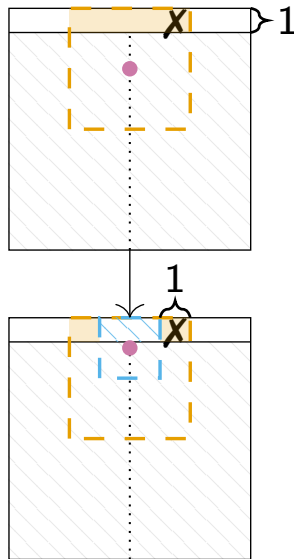
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- 1 more probe,  $\mathcal{O}(1)$  distance
- $\leq \delta_{\min}$  distance to reach POI...
- $D(n) \leq 2\delta_{\min} + \mathcal{O}(1)$



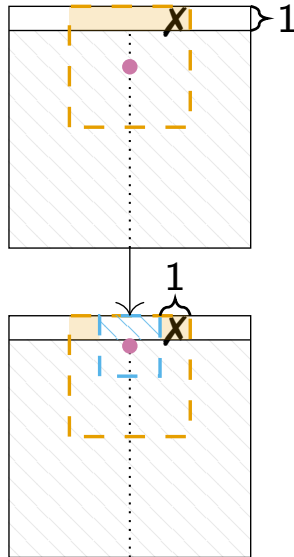
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# Rectilinear Strategies: CBS – 2nd Phase

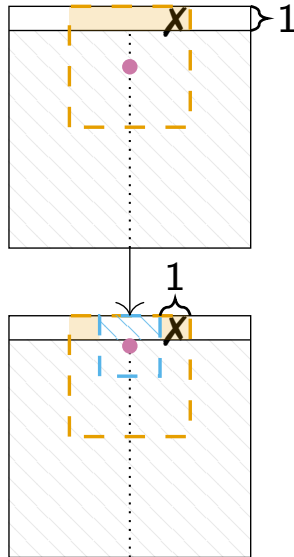
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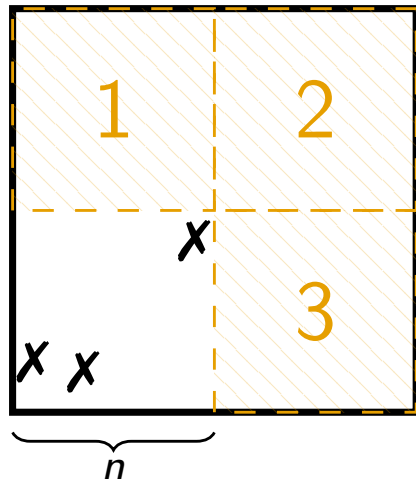
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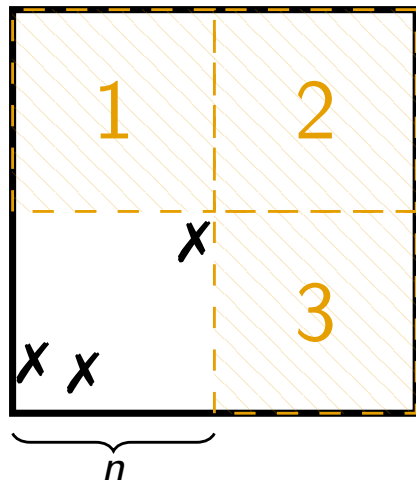
# Rectilinear Strategies: Higher dimensions?

- Is 2D the limit?



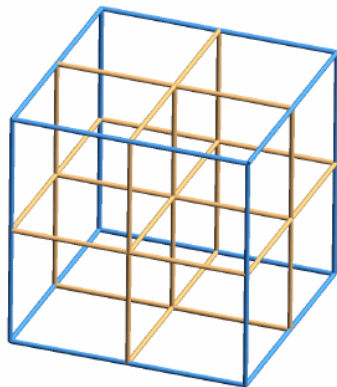
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- Is 2D the limit?
- Recall: Quadrant algorithm



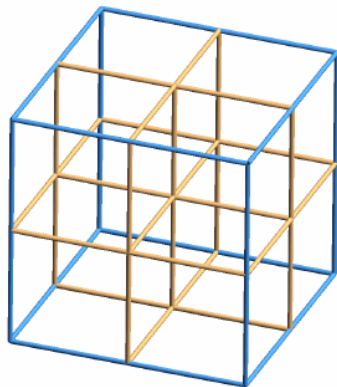
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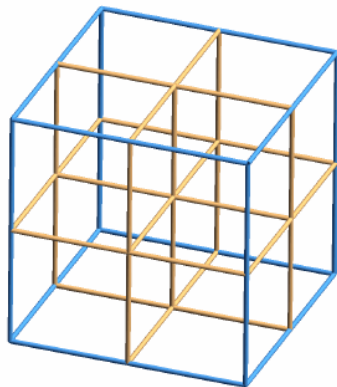
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- $P(n) \leq 7 \lceil \log n \rceil$



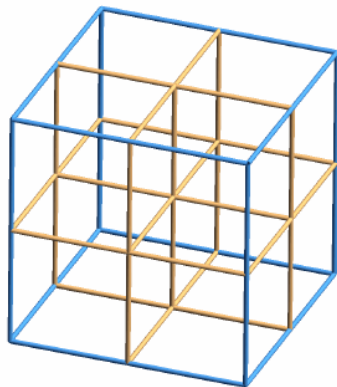
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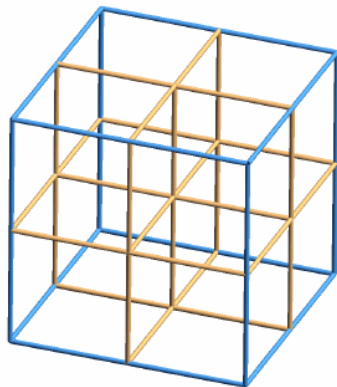
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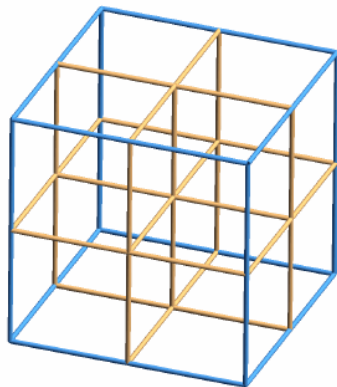
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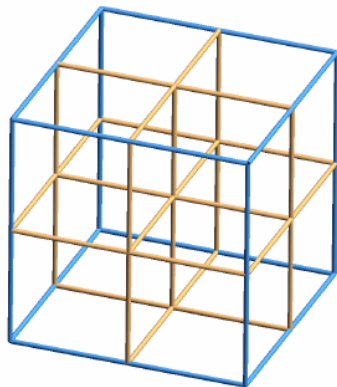
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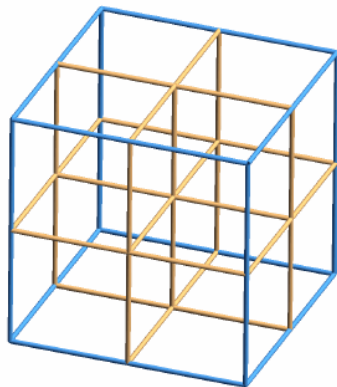
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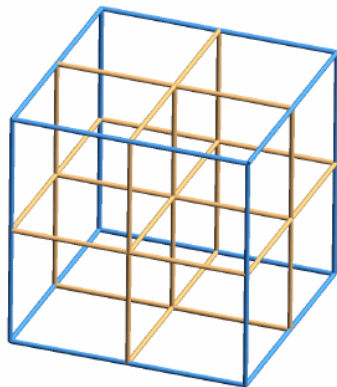
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- Terrible distance  $> 2^k n$
- Excellent responses,  $\leq \lceil \log n \rceil$

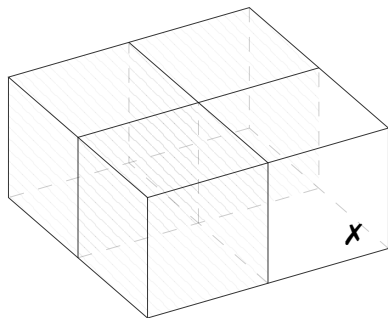


# Rectilinear Strategies: 3D Domino

- How about 3D domino?

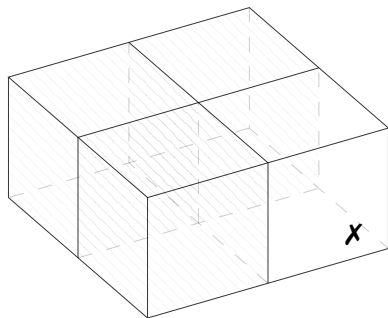
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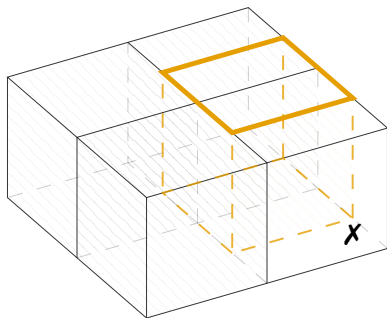
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# Rectilinear Strategies: 3D Domino

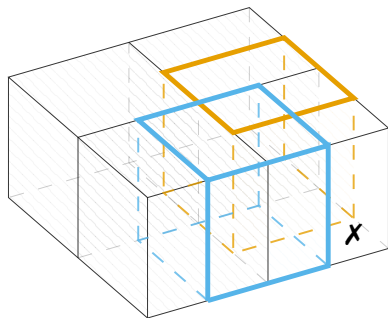
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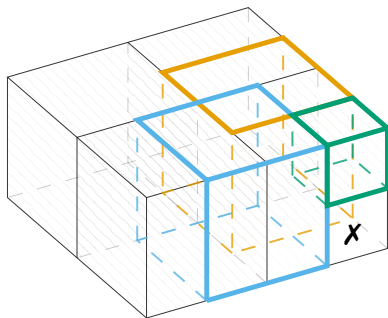
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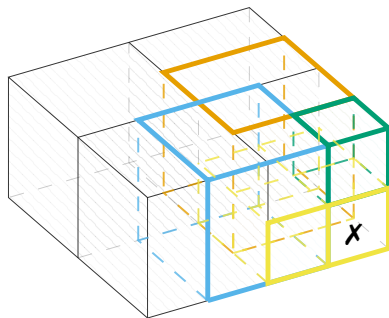
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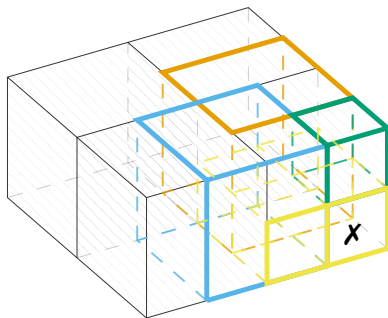
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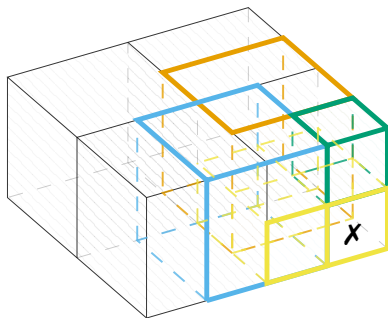
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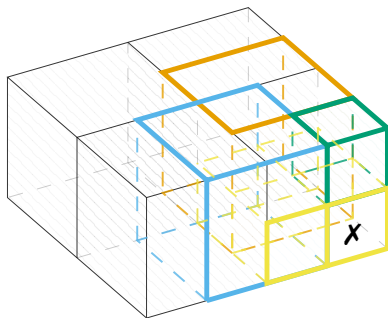
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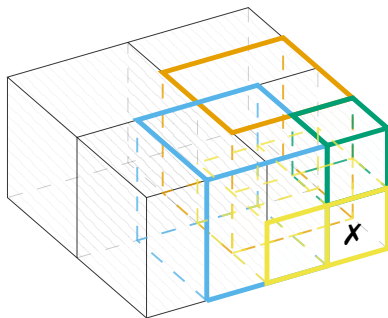
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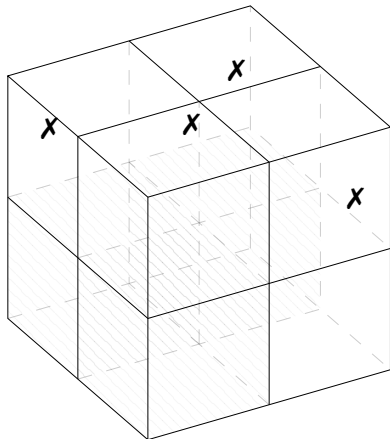
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# Rectilinear Strategies: 3D Domino

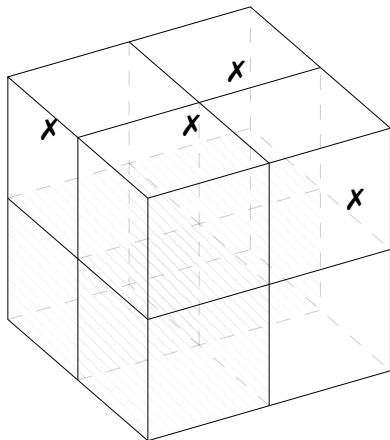
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- $P(n) = 3\lceil \log n \rceil$
- How to reach 4-domino?
- May take more than 4 probes...





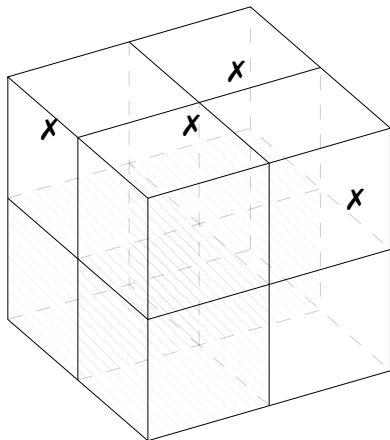
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- How to reach 4-domino?
- May take more than 4 probes...
- If 4th probe always succeeds...



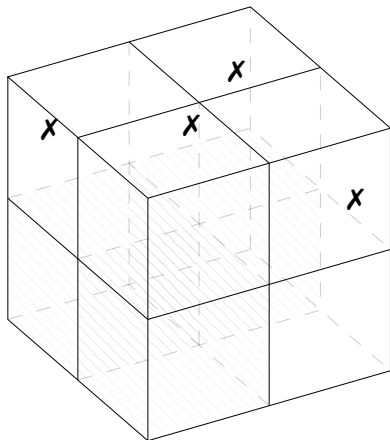
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- Probes halve volume – optimal
- $P(n) = 3\lceil \log n \rceil$
- How to reach 4-domino?
- May take more than 4 probes...
- If 4th probe always succeeds...
- $P(n) \leq 4\lceil \log n \rceil$  – not optimal!

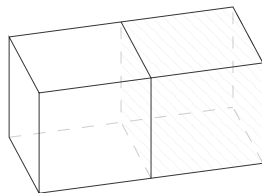


# Rectilinear Strategies: 3D Domino – cont'd

- Idea: Intermediate configuration

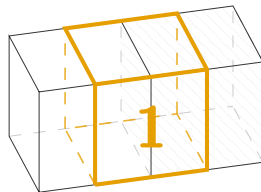
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- Idea: Intermediate configuration
- 3D 2-domino:  $2n \times n \times n$



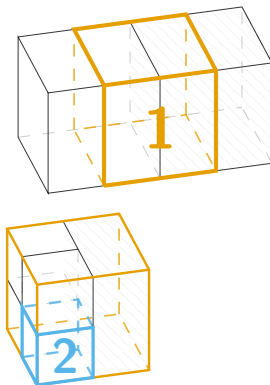
# Rectilinear Strategies: 3D Domino – cont'd

- Idea: Intermediate configuration
- 3D 2-domino:  $2n \times n \times n$
- ① Probe center



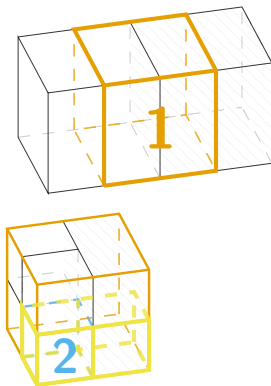
# Rectilinear Strategies: 3D Domino – cont'd

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- ② Probe remaining 'quadrant'



# Rectilinear Strategies: 3D Domino – cont'd

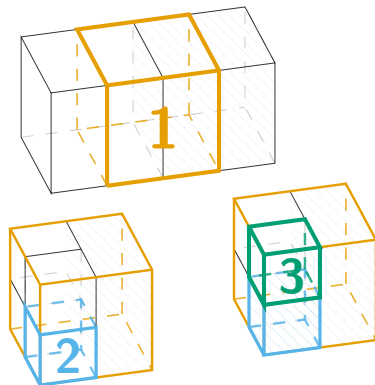
- Idea: Intermediate configuration
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  - ⓐ If success... 2-domino!





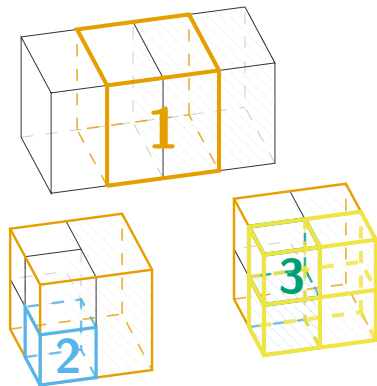
# Rectilinear Strategies: 3D Domino – cont'd

- Idea: Intermediate configuration
- 3D 2-domino:  $2n \times n \times n$
- ① Probe center
- ② Probe remaining 'quadrant'
  - a If success... 2-domino!
  - b Otherwise, more probes...



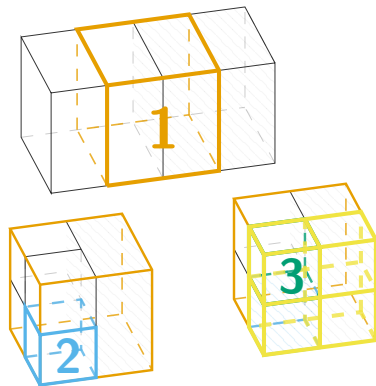
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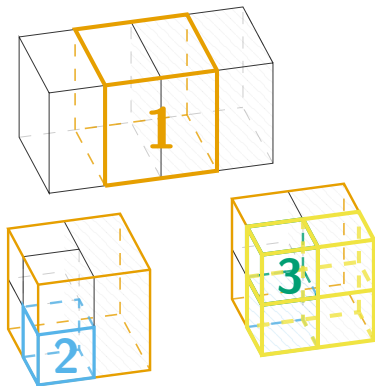
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- $P(n) \leq 3\lceil \log n \rceil + 4$



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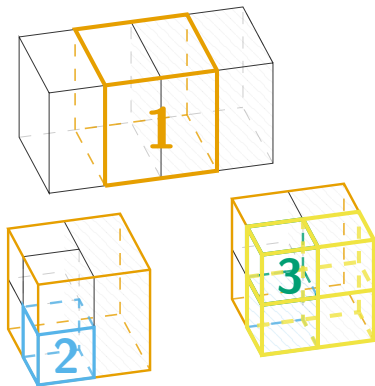
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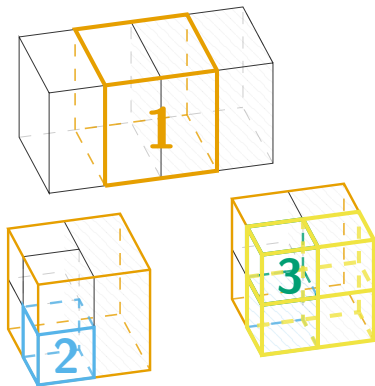
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- Great but...

- Doesn't generalize further ☹️

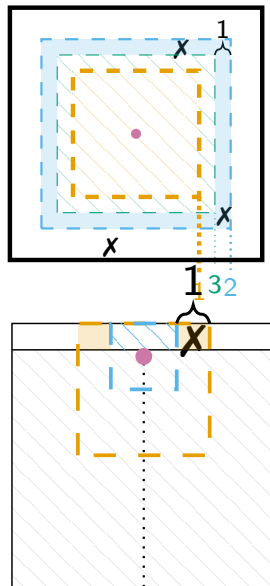


# Rectilinear Strategies: Generalized CBS

- Good general algorithm?

# Rectilinear Strategies: Generalized CBS

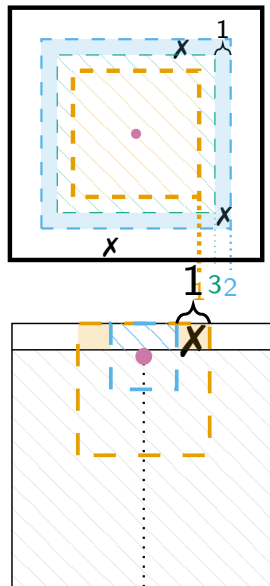
- Good general algorithm?
- Recall CBS—





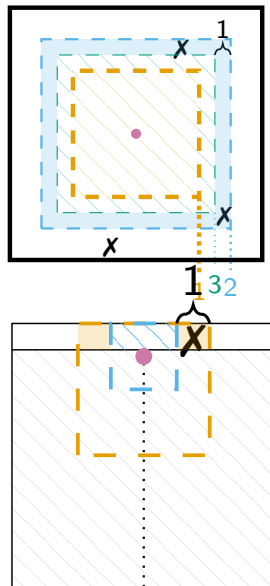
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- Good general algorithm?
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- Binary search per dimension...



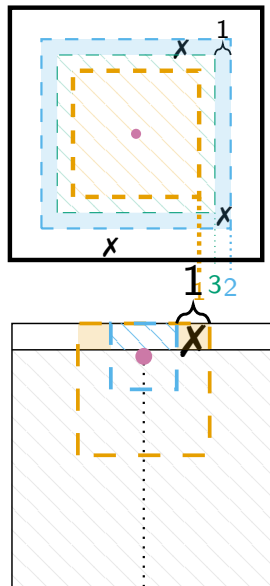
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- Recall CBS—
- Binary search per dimension. . .
- Generalize to  $k$ D?



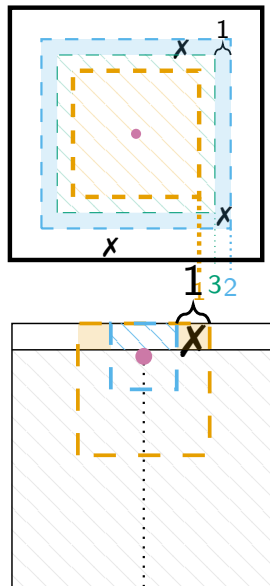
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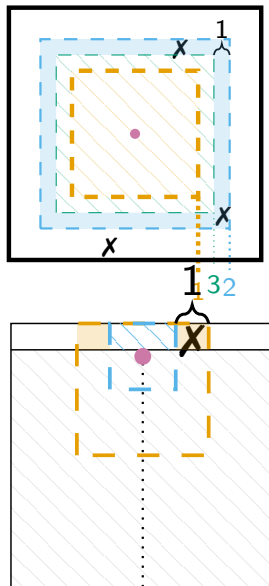
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- Good general algorithm?
- Recall CBS—
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- Generalize to  $k$ D? **yes!**
- Generalized CBS
- $P(n) \leq k \lceil \log n \rceil + g(k)$



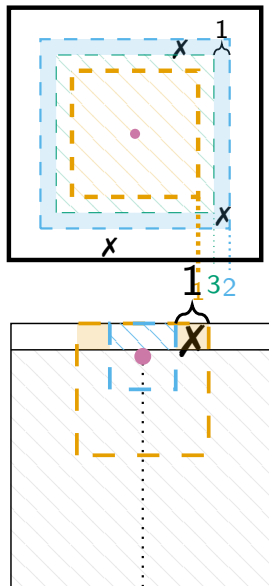
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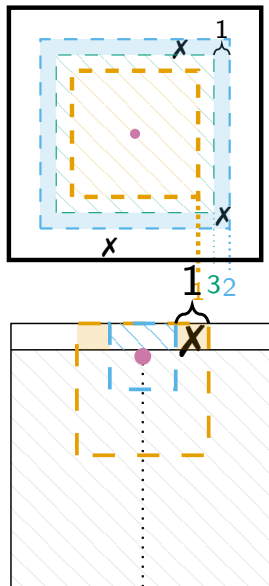
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- How?



# Rectilinear Strategies: Generalized CBS

- Good general algorithm?
- Recall CBS—
- Binary search per dimension. . .
- Generalize to  $k$ D? **yes!**
- Generalized CBS
- $P(n) \leq k \lceil \log n \rceil + g(k)$
- $D(n) \leq k \delta_{\min} + g(k)$
- How?
- Read the full paper :P



# Rectilinear Strategies: Comparing # Probes

$k$	Orthant Algorithm				Generalized CBS Algorithm			
	$\sigma$	Avg	Max	Bound	$\sigma$	Avg	Max	Bound
1D	<b>0.00</b>	<b>0.95</b>	<b>0.95</b>	<b>1.00</b>	0.00	<b>0.95</b>	1.00	<b>1.00</b>
2D	0.18	2.14	2.85	3.00	<b>0.03</b>	1.93	2.00	2.15/2.25
3D	0.46	4.16	6.35	7.00	<b>0.07</b>	2.96	3.10	3.40/4.10
4D	0.98	8.02	13.2	15.0	<b>0.10</b>	<b>4.00</b>	<b>4.25</b>	<b>4.75/7.75</b>
5D	2.00	15.6	26.1	31.0	<b>0.14</b>	<b>5.06</b>	<b>5.40</b>	<b>6.20/16.8</b>
6D	4.02	30.9	50.6	63.0	<b>0.17</b>	<b>6.15</b>	<b>6.65</b>	<b>7.75/42.0</b>
7D	8.05	61.3	103	127	<b>0.21</b>	<b>7.26</b>	<b>7.95</b>	<b>9.40/116</b>
8D	16.1	122	209	<b>255</b>	<b>0.25</b>	<b>8.40</b>	<b>9.30</b>	11.2/336

$k$	Domino Algorithms			
	$\sigma$	Avg	Max	Bound
2D	0.04	<b>1.92</b>	<b>1.95</b>	<b>2.05</b>
3D	0.11	<b>2.92</b>	<b>3.05</b>	<b>3.20</b>



# Rectilinear Strategies: Comparing Distance Traveled

$k$	Orthant Algorithm				Generalized CBS Algorithm			
	$\sigma$	Avg	Max	Bound	$\sigma$	Avg	Max	Bound
1D	$\sim 10^4$	27.8	$\sim 10^7$	$\sim 10^6$	<b>0.00</b>	<b>1.00</b>	<b>1.00</b>	<b>1.00</b>
2D	17.1	8.00	$\sim 10^4$	$\sim 10^6$	<b>0.29</b>	<b>1.50</b>	<b>2.00</b>	<b>2.00</b>
3D	8.14	12.0	$\sim 10^3$	$\sim 10^7$	<b>0.41</b>	<b>2.00</b>	<b>3.00</b>	<b>3.00</b>
4D	10.4	21.3	$\sim 10^3$	$\sim 10^7$	<b>0.50</b>	<b>2.50</b>	<b>3.99</b>	<b>4.00</b>
5D	16.7	40.0	$\sim 10^3$	$\sim 10^7$	<b>0.58</b>	<b>3.00</b>	<b>4.97</b>	<b>5.00</b>
6D	29.7	76.8	$\sim 10^3$	$\sim 10^8$	<b>0.65</b>	<b>3.50</b>	<b>5.94</b>	<b>6.00</b>
7D	55.2	149	$\sim 10^3$	$\sim 10^8$	<b>0.71</b>	<b>4.00</b>	<b>6.86</b>	<b>7.00</b>
8D	105	293	$\sim 10^3$	$\sim 10^8$	<b>0.76</b>	<b>4.50</b>	<b>7.76</b>	<b>8.00</b>

$k$	Domino Algorithms			
	$\sigma$	Avg	Max	Bound
2D	17.2	7.68	$\sim 10^4$	$\sim 10^6$
3D	7.17	11.0	$\sim 10^3$	$\sim 10^7$

# Rectilinear Strategies: Comparing # Responses

$k$	Orthant Algorithm				Generalized CBS Algorithm			
	$\sigma$	Avg	Max	Bound	$\sigma$	Avg	Max	Bound
1D	0.11	0.47	<b>0.95</b>	<b>1.00</b>	<b>0.11</b>	<b>0.47</b>	<b>0.95</b>	1.05
2D	<b>0.09</b>	<b>0.71</b>	<b>0.95</b>	<b>1.00</b>	0.15	0.94	1.75	2.10
3D	<b>0.07</b>	<b>0.83</b>	<b>0.95</b>	<b>1.00</b>	0.18	1.42	2.45	3.15
4D	<b>0.05</b>	<b>0.89</b>	<b>0.95</b>	<b>1.00</b>	0.21	1.88	3.00	4.20
5D	<b>0.04</b>	<b>0.92</b>	<b>0.95</b>	<b>1.00</b>	0.23	2.33	3.50	5.25
6D	<b>0.03</b>	<b>0.94</b>	<b>0.95</b>	<b>1.00</b>	0.25	2.78	4.20	6.30
7D	<b>0.02</b>	<b>0.94</b>	<b>0.95</b>	<b>1.00</b>	0.27	3.22	4.65	7.35
8D	<b>0.01</b>	<b>0.95</b>	<b>0.95</b>	<b>1.00</b>	0.28	3.65	5.25	8.40

$k$	Domino Algorithms			
	$\sigma$	Avg	Max	Bound
2D	0.15	0.93	1.70	2.05
3D	0.18	1.39	2.35	3.20

# Open Problems

- 2D rectilinear algs. work under both  $L_1$  and  $L_\infty$  norms  
Confession: Only  $L_\infty$  works for higher dimensions  
Can we adapt general algs. for  $L_1$ ?
- $g(k)$  constant in generalized CBS algorithm pretty large ( $3^k$ )  
Under reasonable assumptions can reduce to  $k^2$   
Can we do better?
- Distance traveled of  $\delta_{\min}$  instead of  $k\delta_{\min}$ ?
- Mix & match distance metrics
- Other norms?