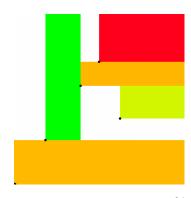
#### Lab Project: Anchored Rectangle Packing

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#### The problem



#### Definition

Given a finite set of points  $P \subset [0,1]^2$ , find a rectangle  $R^p = [p_x, r_x^p) \times [p_y, r_y^p) \subseteq [0,1]^2$  for each  $p \in P$  such that  $R^p \cap R^q = \varnothing$  for all  $p \neq q \in P$ . Area:  $\sum_{p \in P} (r_x^p - p_x) (r_y^p - p_y)$ 

Q: Can you always cover 50% if  $(0,0) \in P$ ?

# Why 50% ?

(interactive)

#### Contents

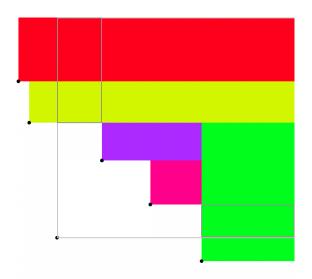
Algorithmic Approach

Optimal Algorithm through Dynamic Programming

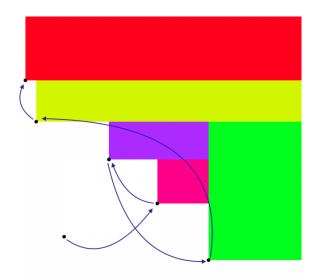
Two Heuristics: TILEPACKING and GREEDYPACKING

Different Greedys

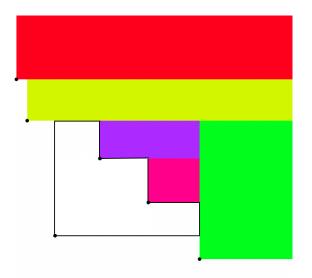
# Ordering



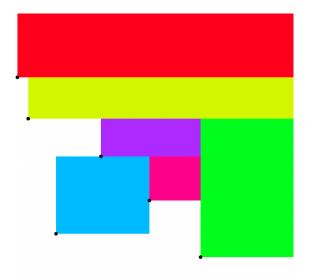
# Ordering



#### Tile



# **Greedy Choices**



#### Basic Algorithm

```
\label{eq:covered} \begin{split} & \text{covered} := 0 \\ & \text{For all permutations } \pi \text{ of } P, \\ & \text{R} := \text{pack rectangles greedily in order } \pi \\ & \text{covered} := \max(\text{s, coverage}(\text{R})) \\ & \text{return covered} \end{split}
```

### **Dynamic Programming**

If  $\pi=\pi',x$  is the optimal permutation, then  $\pi'$  is optimal for  $P\backslash\{x\}$ .

Idea: Inductively compute  $\pi$  for all subsets of P.

Held-Karp Dynamic Programming Solver for TSP  $O(2^n n \log n)$ .

#### Improving it with Heuristics

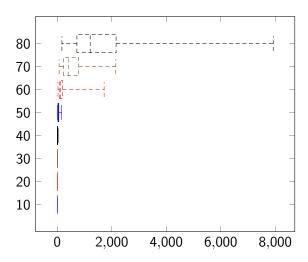


Figure: Time (ms) for the Optimal Algorithm on n uniformly distributed points; 100 instances

#### **TilePacking**

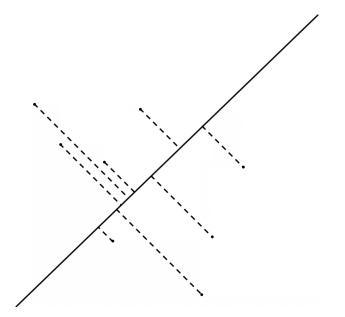
Choose a permutation through a sweep and pack accordingly

Easy to implement in time  $O(n \log n)$ 

Can not cover more than 43% in some instances

# TilePacking: Points

# TilePacking: Permutation



### Problem with TilePacking in Practice

(interactive)

#### Performance in Theory

GreedyPacking is no better in theory

Reduction by adding 2 points  $(x - \epsilon, y), (x, y - \epsilon)$  for each  $(x, y) \in P$ 

Corollary: GreedyPacking can not reach 50% either!

#### Performance in Practice

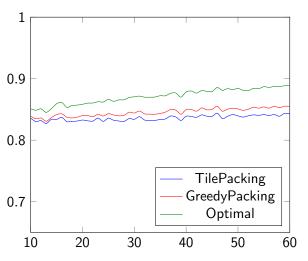


Figure: Average coverage of the algorithms on n uniformly random points; 100 instances

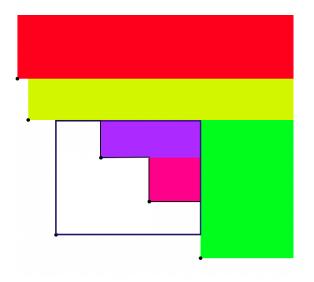
#### GreedyPacking

Best previous algorithm:  $O(n^2 \log n)$  time,  $O(n^2)$  space

Implemented:  $O(n^2)$  time, O(n) space

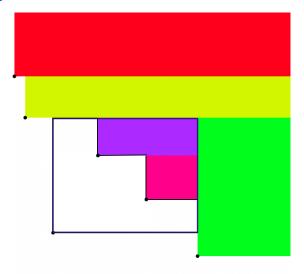
Described:  $O(n \log^2 n + k)$  time,  $O(n \log^2 n)$  space

### Find Tile Rectangle



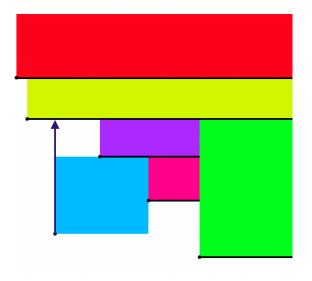
O(n) time, no extra storage

#### Make greedy choice



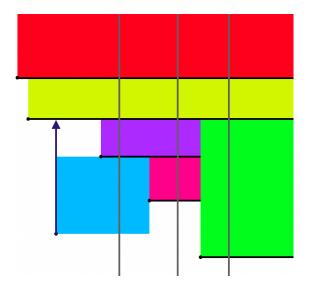
O(n) time, no extra storage  $O(\log^2 n + k)$  time for k points in the tile rect, using a range tree

### Find Tile Rectangle (better)



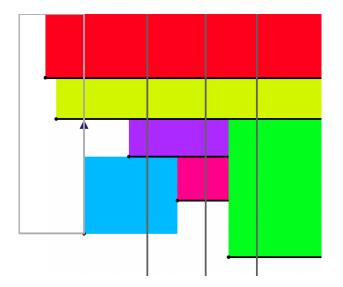
 $O(\log^2 n)$  time using interval and priority search tree

#### Find Tile Rectangle (better)



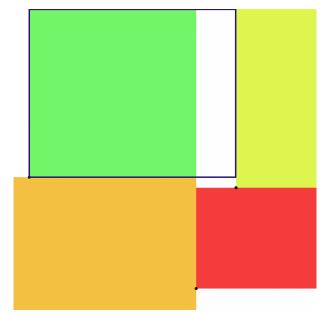
 $O(\log^2 n)$  time using interval and priority search tree

#### Find Tile Rectangle (better)



 $O(\log^2 n)$  time using interval and priority search tree

### Solves the TilePacking Problem



#### GreedyPacking: runtime

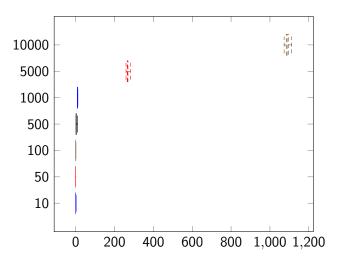
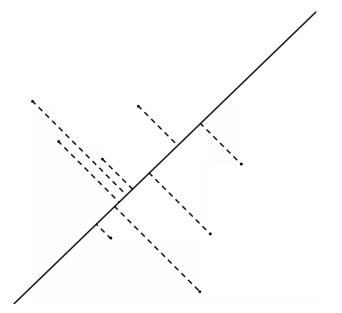
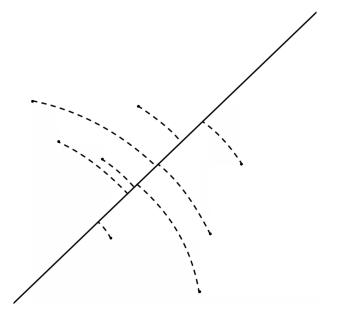


Figure: Time (ms) for the Greedy Algorithm on n uniformly distributed points; 100 instances

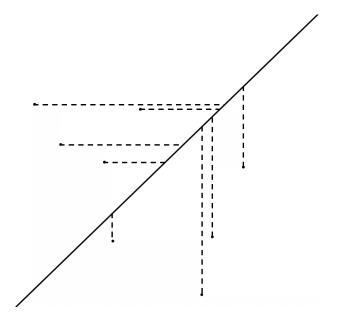
# GreedyPacking: Manhattan



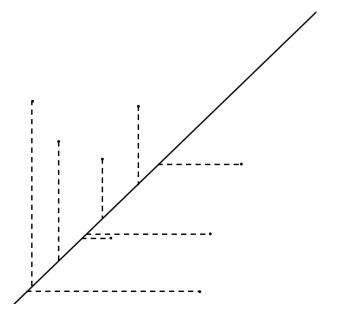
# GreedyPacking: Euclidean



# GreedyPacking: Infinity



### GreedyPacking: Minus Infinity



#### Norms in Practice

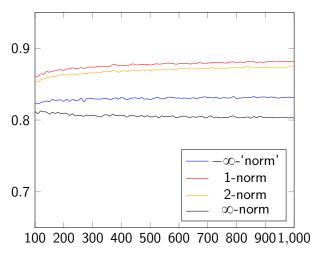


Figure: Average coverage of the greedy algorithm on n uniformly random points; 100 instances

### Why are other norms interesting?

No norm is better than any other!

Damerius found no counter-example for the  $(-)\infty$ -norms

#### Interactive

### Questions?