

# A 2D Bin Packing Problem

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# Introduction

A quick brief to understand  
what we're dealing with

# Packages

A large number of items  
(rectangles) varying in size,  
and we need to deliver them  
all to our customers.



# Trucks

To deliver those packages, vehicles are needed. Each truck carries a container, which also varies in size. In addition, an operation cost is applied at random to every truck.

**\$300 - 4x4**

Truck 1

**\$500 - 3x5**

Truck 2

**\$850 - 1x2**

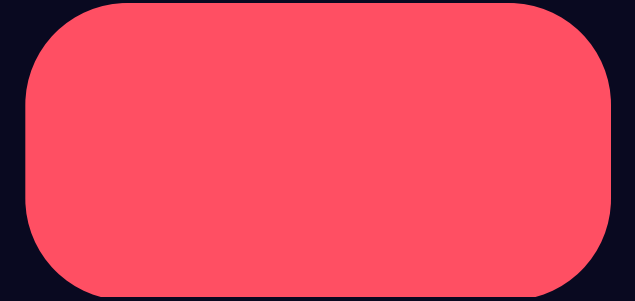
Truck 3

# Objective

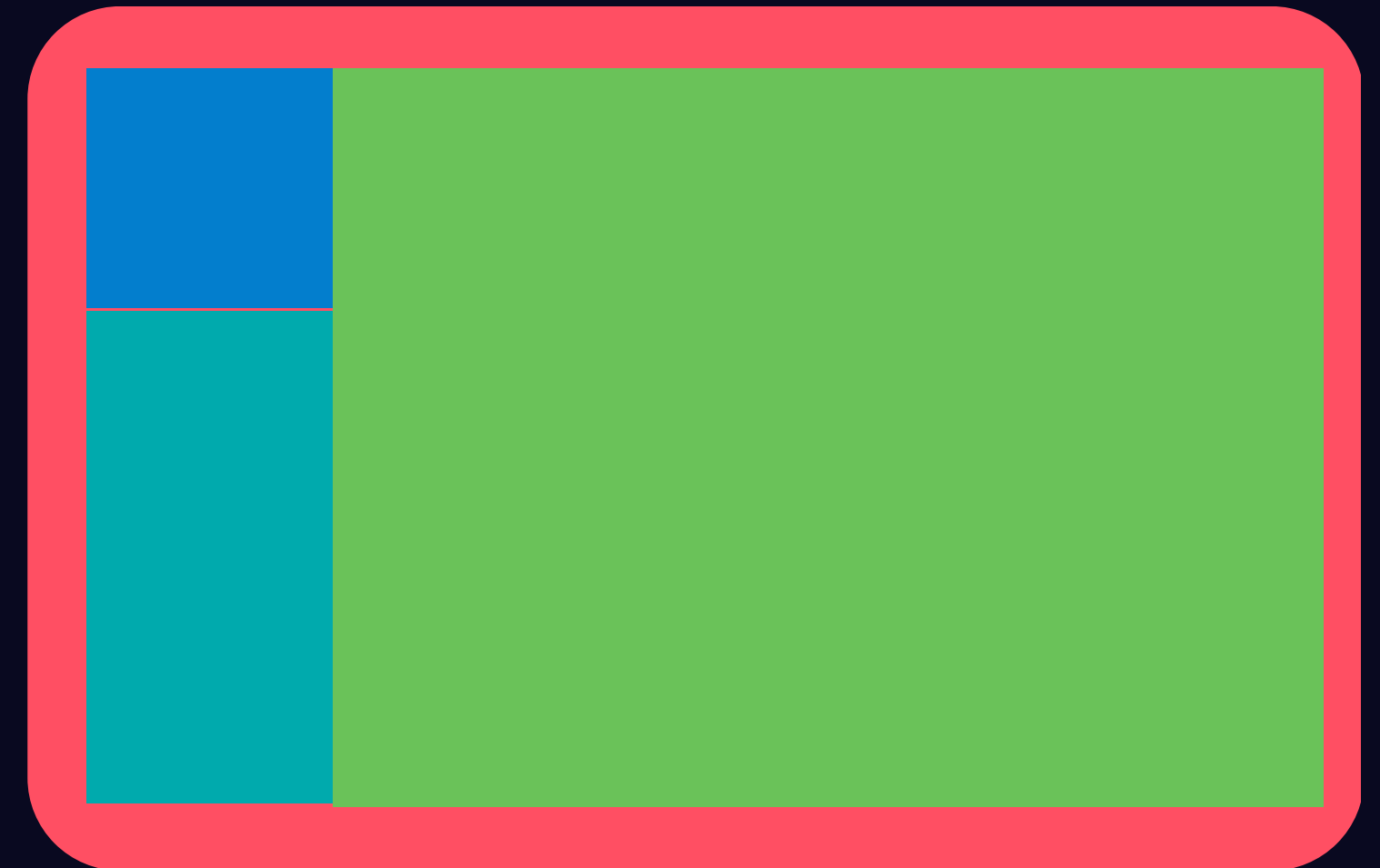
The goal here is to minimize the cost of operation while having all packages delivered.



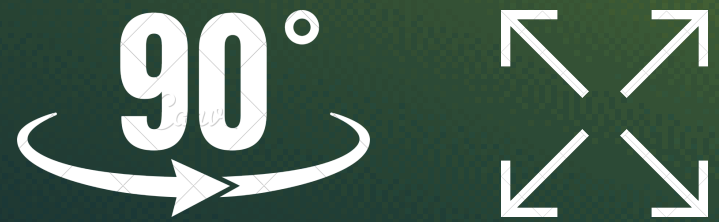
Unused because of  
ridiculus cost



**Total cost: \$800**



# Some Remarks



All packages  
can be rotated  
at an  $90^\circ$  angle  
so that all items  
in a truck could  
fit in it  
orthogonally



Each package has a size and must  
be in one of the trucks



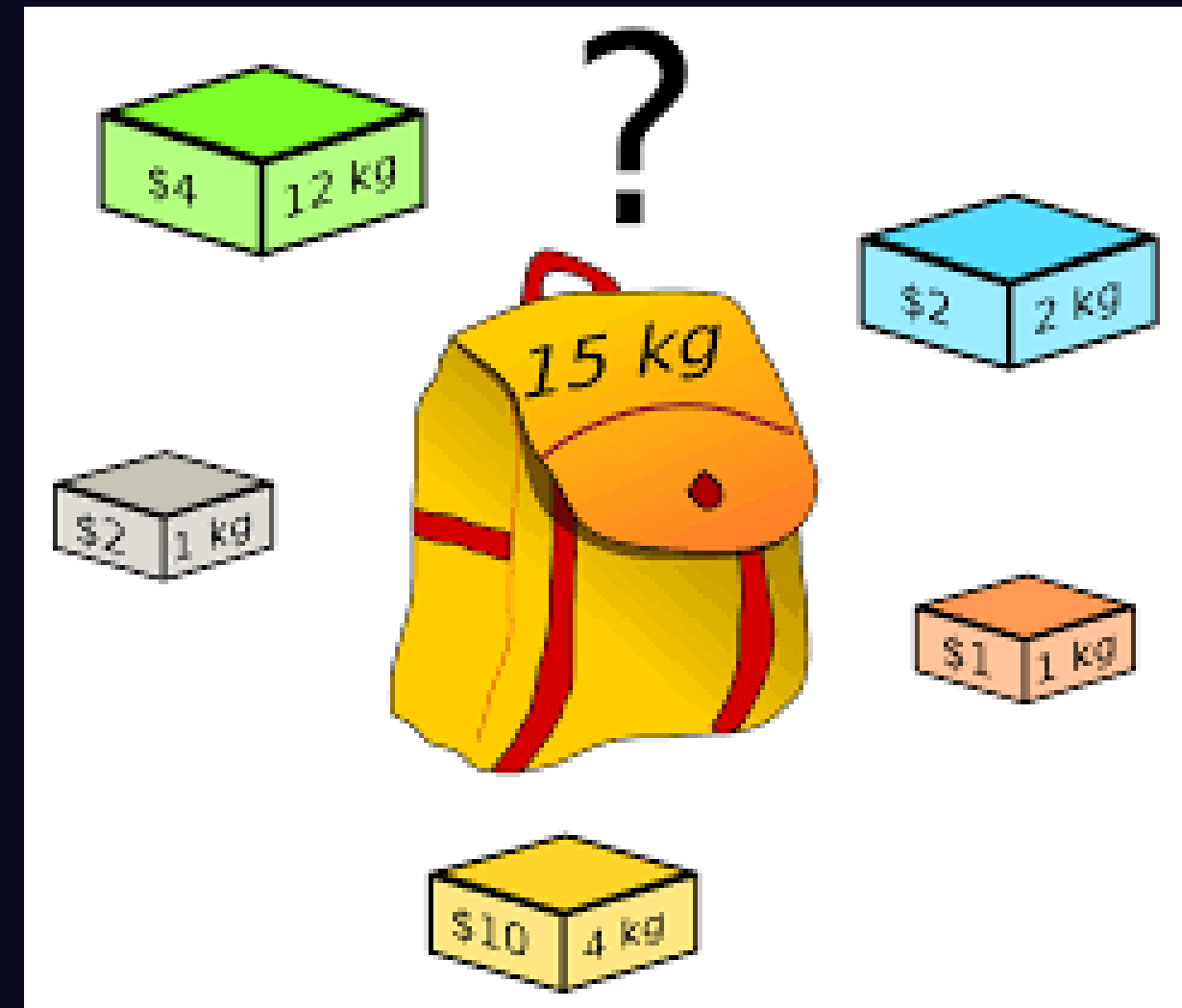
Each truck has a storage capacity  
and operation cost



Minimize the total cost of trucks  
used in delivery

# Fun Fact!

This problem is a variant of the 1-dimensional "Knapsack Problem", where items has a "weight/size" and "value".





# Data Generation

# Set a package count for each file

74 files are generated with increasing number of packages

- 50 **basic and intermediate** files: 5-54 packages with step 1  
i.e. 5, 6, 7,...
- 10 **hard** files: 60-330 packages with step 30  
i.e. 60, 90, 120,...
- 14 **very hard** files: 350-1000 packages with step 50  
i.e. 350, 400, 450,...
- Along with 1 sample file, creating a data set of 75 files in total.

# Data Generating Algorithm

## Step 1

Randomize  
packages

## Step 2

Randomly load  
packages in  
trucks

## Step 3

Shrink  
trucks' sizes

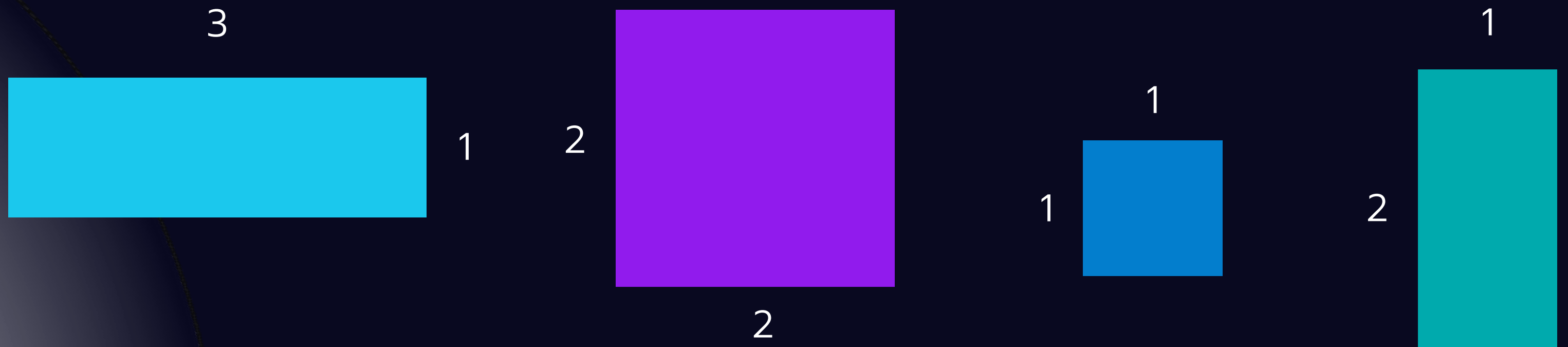
## Step 4

Randomly  
generate a few  
more trucks and  
then assign cost  
for all trucks

# Randomize Packages

Each package has sizes  $a \times b$ , in which  $a$  and  $b$  can be any integer from 1 to 10.

$n$  iterations are needed to obtain  $n$  packages.



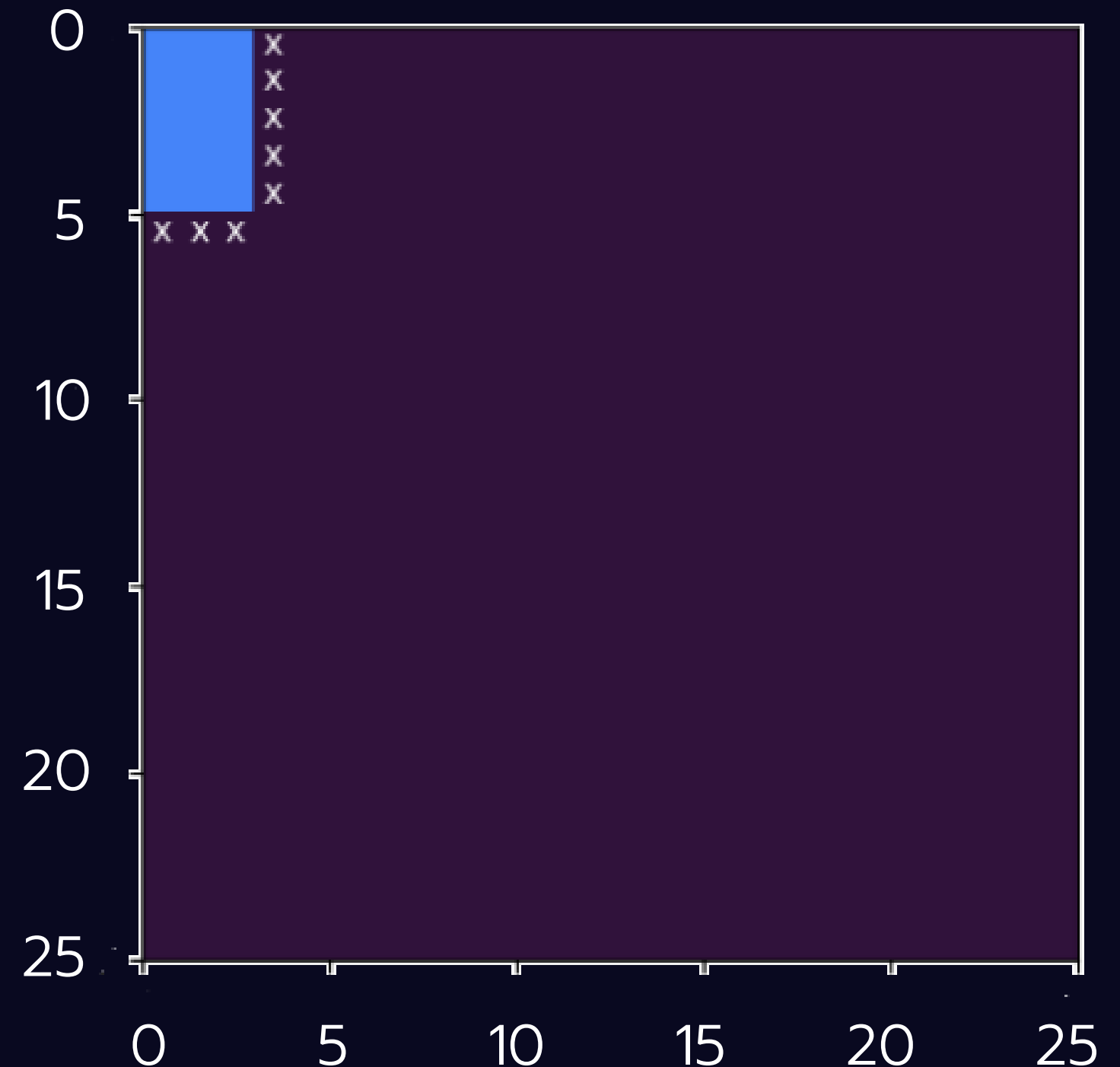
# Randomly put packages in trucks

Out of  $n$  packages, pick from 2 to 5 at random.

Initialize a 30x30 truck.

Put the first package in the upper left corner.

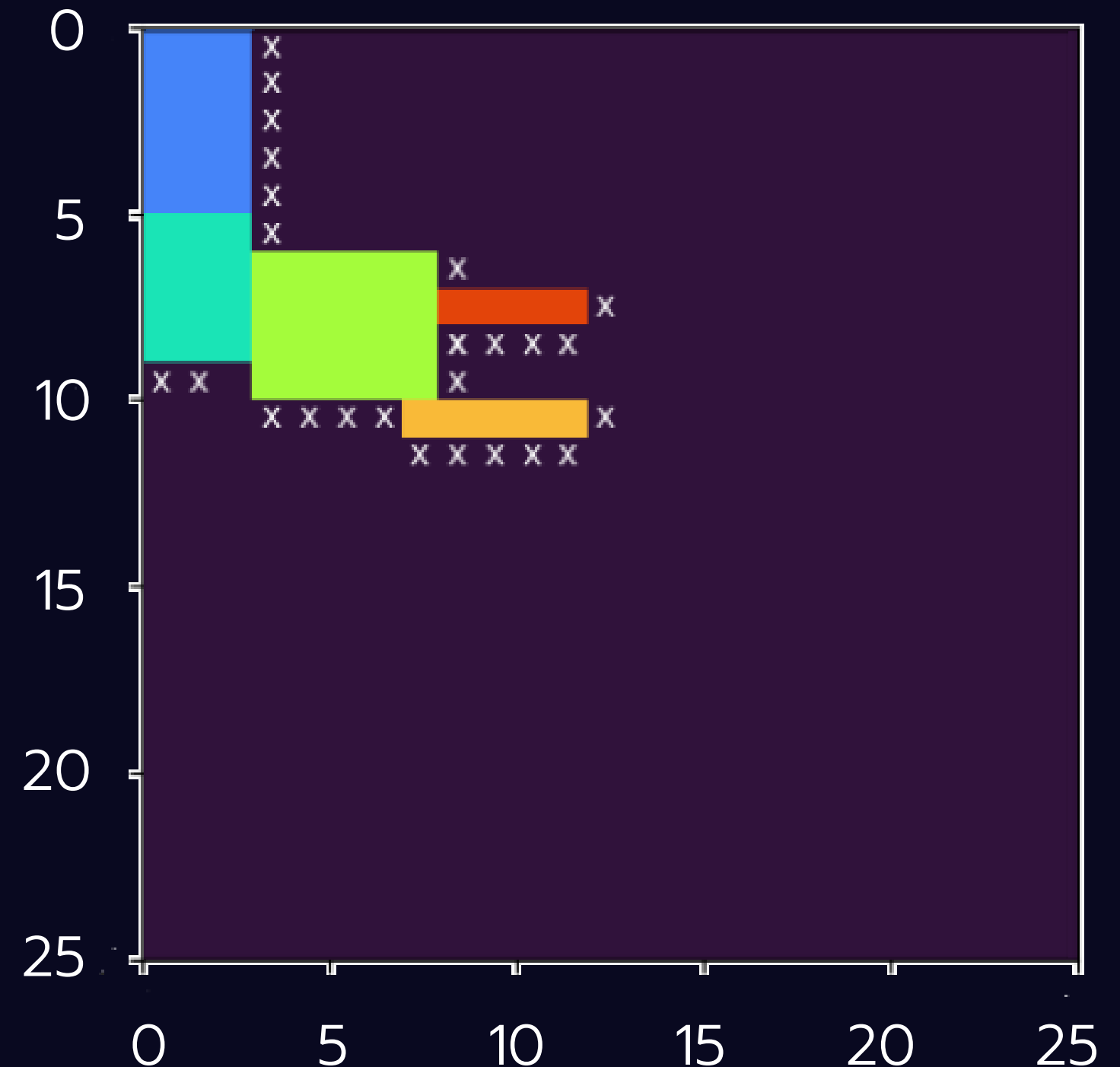
Mark all the available spaces on the right or bottom of the package.



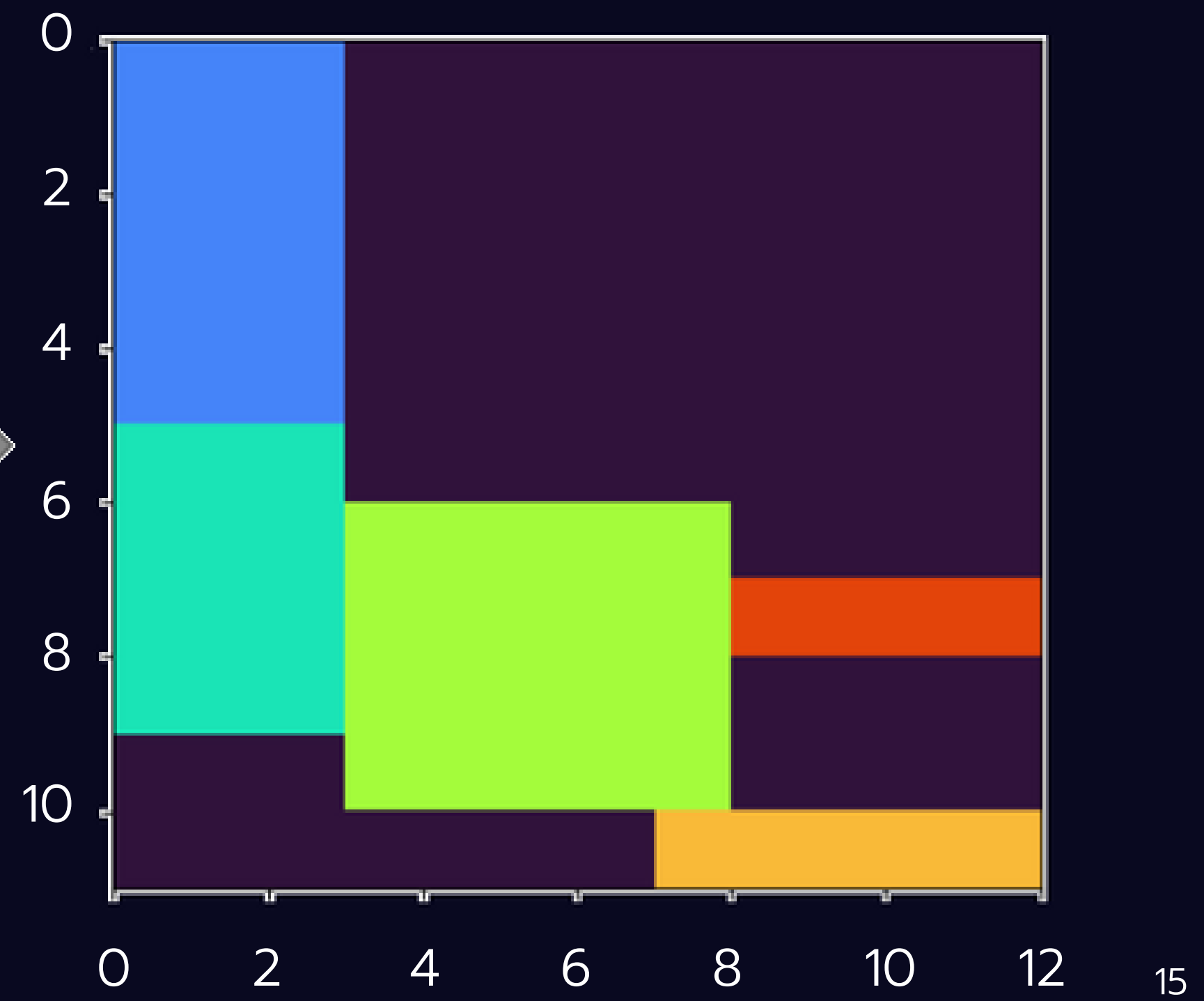
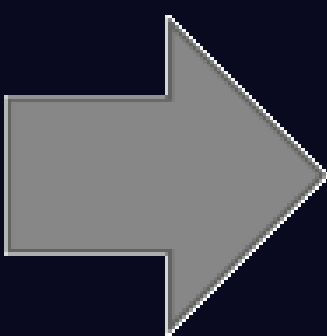
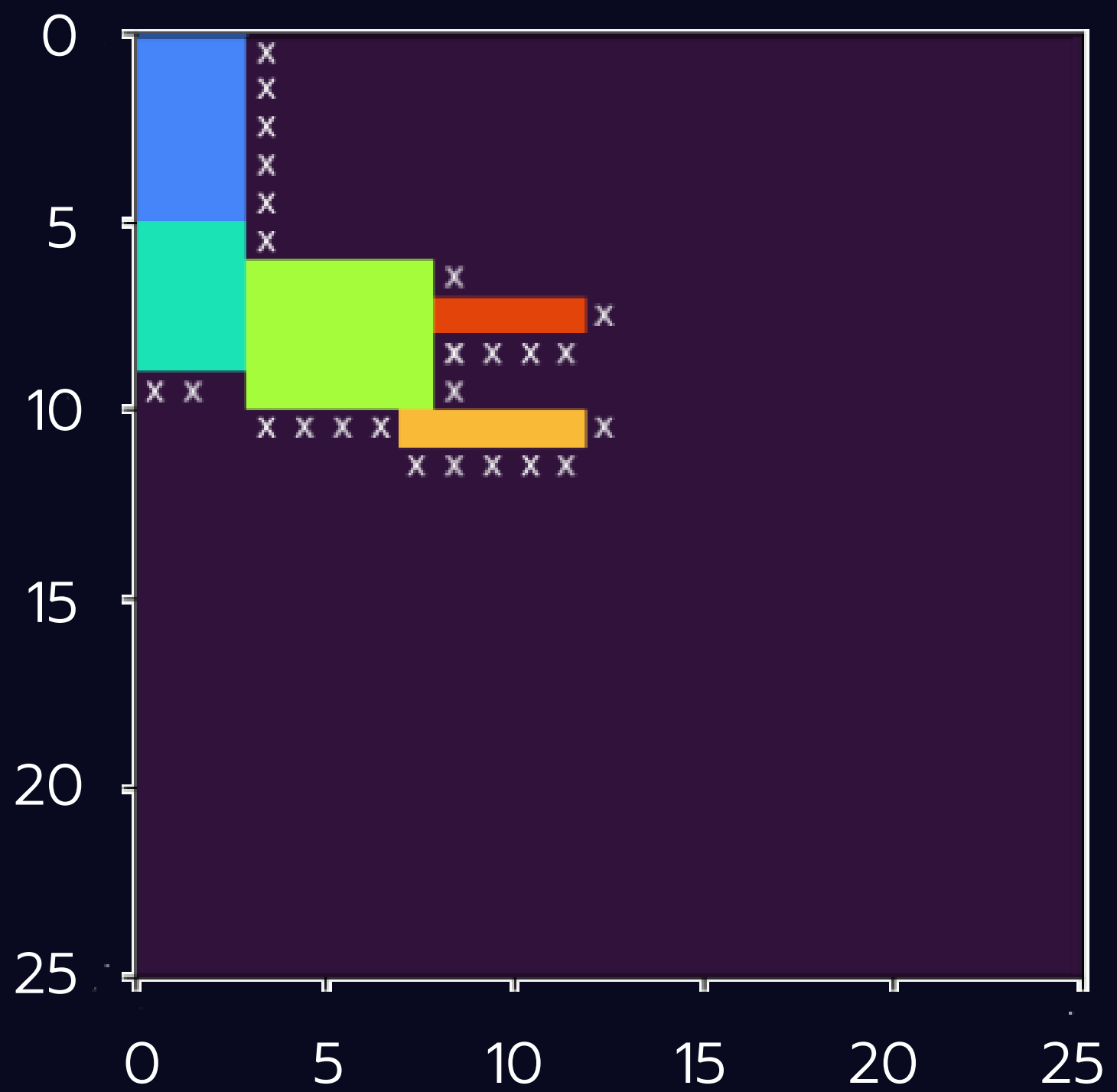
# Randomly put packages in trucks

Choose one out of the remaining package(s) and randomly put it in the truck.

Repeat the above step until no packages remain.



# Shrink trucks' sizes



# **Randomly generate a few more trucks and then assign cost for all trucks**

**Generate a few more trucks with their dimensions varying from 1 to 25  
(i.e. 1x18, 21x12, 20x14, 25x25, etc.)**

**Assign cost from 100 to 1000 with step 50  
(i.e. 100, 150, 200, ..., 1000)**



# Building CP and MIP models

# Denotation

$R = \{1, 2, \dots, n\}$  is the set of given packages.  
Item  $i$  has width  $w_i$  and height  $h_i$ .

$B = \{1, 2, \dots, m\}$  is the set of available trucks.  
Container  $k$  has width  $w_k$ , height  $h_k$  and cost  $c_k$ .

# Variables

$o_i \in \{0, 1\}$  represents the orientation of package  $i$ .

$l_i, r_i, t_i, b_i$  are left, right, top and bottom coordinates of package  $i$ .

Binary variable  $u_k$  is 1 if truck  $k$  is used.

Binary variable  $p_{ik}$  is 1 if package  $i$  is placed in truck  $k$ .

$y_i = k \in \{1, \dots, m\}$ , i.e. package  $i$  is placed in truck  $k$ .

# CP Model - Constraint

$$o_i = 0 \Rightarrow r_i = l_i + w_i \wedge t_i = b_i + h_i \quad \forall i \in R \quad (1)$$

$$o_i = 1 \Rightarrow r_i = l_i + h_i \wedge t_i = b_i + w_i \quad \forall i \in R \quad (2)$$

$$y_i = y_j \Rightarrow r_i \leq l_j \vee r_j \leq l_i \vee t_i \leq b_j \vee t_j \leq b_i \quad \forall i, j \in R, i < j \quad (3)$$

$$y_i = k \Rightarrow r_i \leq W_i \wedge t_i \leq H_i \quad \forall i, j \in R, k \in B \quad (4)$$

$$p_{ik} = 1 \Rightarrow y_i = k \quad \forall i \in R, k \in B \quad (5)$$

$$\sum_{i=1}^n p_{ik} \geq 1 \Rightarrow u_k = 1 \quad \forall k \in B \quad (6)$$

# CP Model - Constraint

- (1) If an item doesn't rotate, its  $\text{right} = \text{its left} + \text{its width}$  and its  $\text{top} = \text{its bottom} + \text{its height}$
- (2) If the item rotates then its  $\text{right} = \text{its left} + \text{its height}$  and its  $\text{top} = \text{its bottom} + \text{its width}$
- (3) If two items are placed in the same bin, they can't overlap each other
- (4) If one item is placed in a bin then its right and top coordinates can't exceed the bin
- (5) Item  $i$  is placed in bin  $k$
- (6) Bin  $k$  is used when at least one item is placed in it

# MIP Model – Constraint

$$o_i = 0 \Rightarrow r_i = l_i + w_i \wedge t_i = b_i + h_i \quad (1); \quad o_i = 1 \Rightarrow r_i = l_i + h_i \wedge t_i = b_i + w_i \quad (2)$$

$$\text{To MIP: } r_i = l_i + w_i * (1 - o_i) + h_i * o_i; \quad t_i = b_i + h_i * (1 - o_i) + w_i * o_i$$

$$y_i = y_j \Rightarrow r_i \leq l_j \vee r_j \leq l_i \vee t_i \leq b_j \vee t_j \leq b_i \quad (3)$$

$$\text{To MIP: } r_i \leq M * (1 - x_1) + l_j; \quad r_j \leq M * (1 - x_2) + l_i;$$

$$t_i \leq M * (1 - x_3) + b_j; \quad t_j \leq M * (1 - x_4) + b_i;$$

$$x_1 + x_2 + x_3 + x_4 + (1 - x_0) * M \geq 1;$$

$$x_1 + x_2 + x_3 + x_4 \leq x_0 * M; \quad b = 1 - x_0;$$

$$y_j + x_0 \leq y_i + M * x_{01}; \quad y_i + x_0 \leq y_j + M * x_{02}; \quad x_{01} + x_{02} \leq x_0;$$

# MIP Model - Constraint

$$y_i = k \Rightarrow r_i \leq W_i \wedge t_i \leq H_i \quad (4)$$

$$\text{To MIP: } k + x_{00} \leq y_i + M * x_{01}; y_i + x_{00} \leq k + M * x_{02}; x_{01} + x_{02} = x_{00};$$

$$l_i \leq w_k + x_{00} * M; r_i \leq w_k + x_{00} * M; t_i \leq h_k + x_{00} * M; b_i \leq h_k + x_{00} * M;$$

$$p_{ik} = 1 \Rightarrow y_i = k; \text{To MIP: } y_i = k * p_{ik} \quad (5)$$

$$\sum_{i=1}^n p_{ik} \geq 1 \Rightarrow u_k = 1 \quad (6)$$

$$\text{To MIP: } S = \sum_{i=1}^n p_{ik}$$

$$S \leq (1 - e) * M; S + e * M \geq 1; u_k = 1 - e;$$

# Heuristics



# The Brute-force Algorithm

First step: Sort

Trucks are sorted based on their cost/area ratio, in ascending order.

Packages are sorted based on their size, in descending order.

This is done in order to maximize efficiency by prioritizing cost-efficient trucks and largest packages.

# The Brute-force Approach

Second step: Best-fit

From the sorted list, initialize the trucks, each containing no packages.

Iterate through the truck list for each package. If the package fits in the container, break the truck loop and check the next item.

# The Brute-force Approach

Second step: Best-fit

The larger packages will fit in some of the best trucks, and when the algorithm iterate through the list, some smaller packages can also fit in the best trucks.

Therefore we can make use of the best trucks as much as possible.

# The Brute-force Approach

But why sort the packages?

Top trucks on the list are usually the most spacious ones, and thus leave room for items that take up considerable space.

This also makes the trucks have fewer packages, reducing the backtracking algorithm's running time.

# A Thousand Ways to Pack the Bin - A Practical Approach to Two-Dimensional Rectangle Bin Packing

Jukka Jylänki

February 27, 2010

## Abstract

We review several algorithms that can be used to solve the problem of packing rectangles into two-dimensional finite bins. Most of the presented algorithms have well been studied in literature, but some of the variants are less known and some are apparently regarded as "folklore" and no previous reference is known. Different variants are presented and compared. The main contribution of this survey is an original classification of these variants from the viewpoint of solving the finite bin packing problem. This work focuses on empirical studies on the problem variant where rectangles are placed orthogonally and may be rotated by 90 degrees. Synthetic tests are used as the main benchmark and solving a practical problem of generating texture atlases is used to test the real-world performance of each method. As a related contribution, an original proof concerning the number of maximal orthogonal rectangles inside a rectilinear polygon is presented.

**Keywords:** Two-dimensional bin packing, optimization, heuristic algorithm, on-line algorithm, NP-hard

# Another Approach: Guillotine

# Another Approach: Guillotine

This algorithm is based on a operation called "guillotine split placement". This procedure involves placing a package in the corner of a container, after which the remaining space is split into two disjoint smaller space.

The process of packing several items is then modelled as an iterative application of the above operation.

# Another Approach: Guillotine

How it works

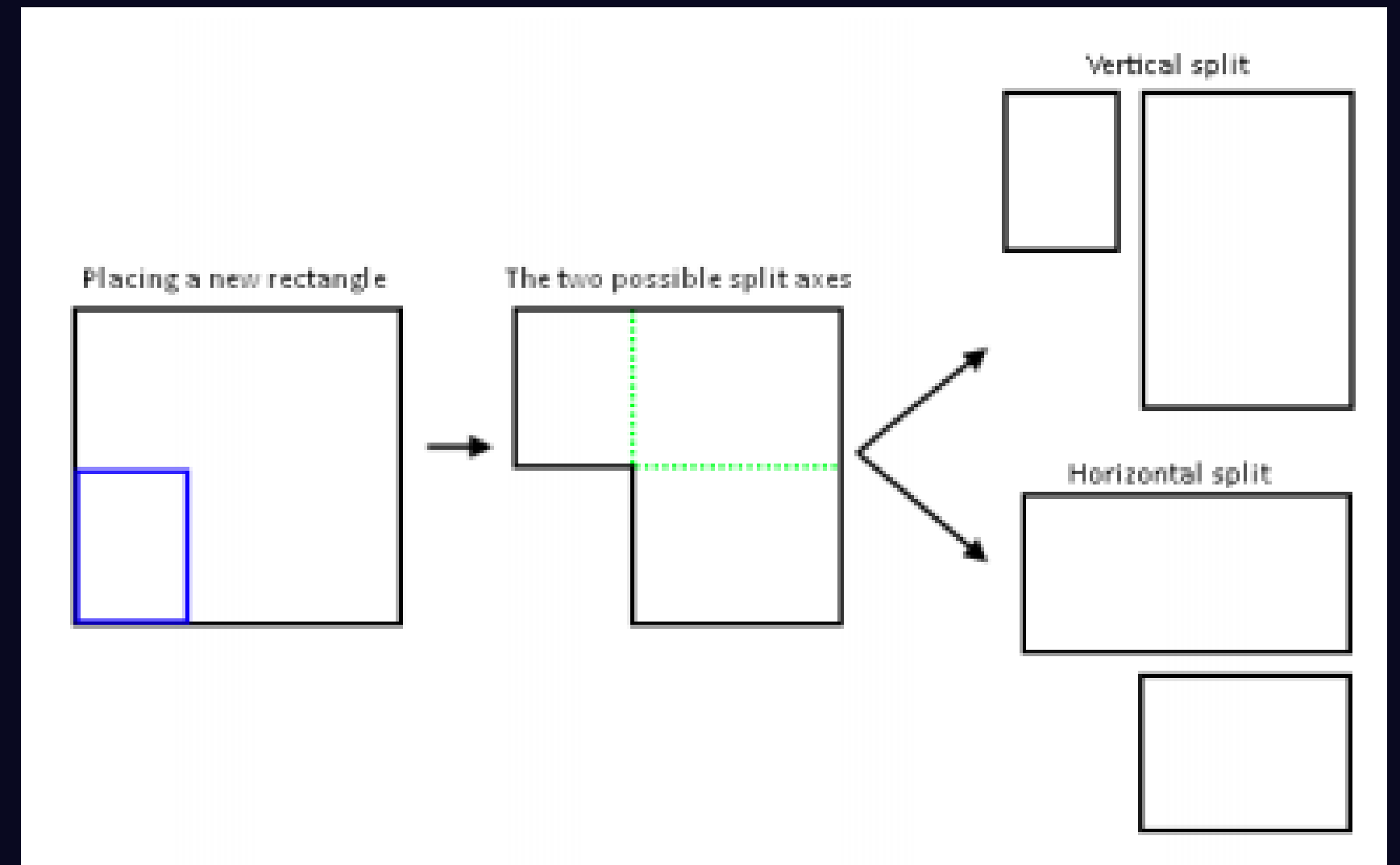
A list of smaller containers  $F = \{F_1, \dots, F_n\}$  (which are pairwise disjoint) representing the free space of the truck is created.

The algorithm starts with a single container  $F = \{F_1\}$ . At each packing step, a free container  $F_i$  is picked at first to place the next package into.

# Another Approach: Guillotine

How it works

The item is placed at the bottom left corner of  $F_i$ , which is then split using the guillotine rule to make two smaller containers  $F'$  and  $F''$ , which replace  $F_i$  in the list of free containers.





# Another Approach: Guillotine

This algorithm is appealing as it monitors the free sections of the truck and never "forgets" any free space.

The downside is that the algorithm only tries to fit items in a single free space  $F_i$ .

# Algorithm Analysis

# REFERENCES

(1): Jukka Jylänki's A Thousand Ways to Pack the Bin - a Practical Approach to Two-Dimensional Rectangle Bin Packing