

# **Programming problems related to chip design**

*B.Tech. Project Report*

*by*

**Akansha Madavi**

**2103104**

*Faculty Advisor*

**Dr. Sharad Sinha**



**Computer Science and Engineering  
Indian Institute of Technology Goa**

## Abstract

In the modern era of electronics and logistics, optimizing space utilization plays a critical role in reducing costs, improving efficiency, and enhancing performance. This project investigates advanced algorithms for efficiently packing smaller boxes into a larger container, reflecting real-world challenges such as chip design and storage optimization. The work is motivated by the need for solutions that can handle not only basic packing tasks but also complex constraints like power consumption, thermal dissipation, and interconnect routing in chip layouts.

The study evaluates and compares multiple algorithmic approaches:

1. **Greedy Placement Algorithm:** A straightforward method focusing on sequential placement to maximize area utilization, suitable for simple scenarios.
2. **Constraint-Driven Algorithm:** An enhanced method integrating power consumption constraints to ensure that adjacent boxes do not exceed specific thresholds, essential for heat-sensitive applications.
3. **Heuristic-Based Approach:** A method that prioritizes larger boxes and considers rotations to minimize gaps and improve initial packing efficiency.
4. **Grid-Based Optimization:** A precise approach leveraging fine-grained grid division to guide placement decisions, handle rotations, and adapt to multiple constraints, achieving the highest utilization in complex scenarios.
5. **Test Case-Based Evaluation:** A comprehensive framework for testing algorithms under diverse real-world conditions, including irregular box sizes, constrained spaces, and overpacking scenarios.

Each algorithm was tested on a 10x10 grid with smaller boxes of varying dimensions, producing metrics such as percentage utilization, packing flexibility, and adaptability to constraints. Results showed the grid-based optimization method achieving the highest utilization (up to 94%), while the constraint-driven approach excelled in scenarios with power and adjacency rules. Test cases highlighted the algorithms' strengths and limitations, providing insights into their suitability for different practical applications.

This research offers significant contributions to industries like chip manufacturing, where efficient layout design reduces silicon costs and enhances performance; and logistics, where maximizing space utilization in storage and transportation can cut expenses and streamline operations. The visualizations provided through this study enable a clear understanding of packing configurations, making the algorithms accessible and practical for diverse use cases.

Future work focuses on incorporating advanced methodologies such as genetic algorithms, simulated annealing, and AI-driven placement strategies to further optimize layouts. Additionally, parallel processing and GUI-based tools will enhance scalability and usability, ensuring these techniques remain relevant in rapidly evolving technological landscapes.

In conclusion, this project bridges the gap between theoretical optimization and practical implementation, delivering robust, scalable, and efficient solutions for real-world challenges in space utilization and layout design.

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# CHAPTER 1 : INTRODUCTION

## 1.1 What is Chip Design?

Chip design is a specialized engineering process focused on creating the physical and functional layout of integrated circuits (ICs), which are the backbone of modern electronic devices. It involves crafting semiconductor chips that house thousands to billions of tiny transistors, logic gates, and other components to perform specific tasks. These chips are used in various applications, from microprocessors in computers to sensors in IoT devices and controllers in vehicles.

**Chip design encompasses multiple stages, including:**

- 1. Specification and Architecture:** Defining the chip's functionality, performance goals, and constraints (e.g., power, area, and speed).
- 2. Logic Design:** Developing the circuit's behavior using logic gates, ensuring it performs as intended.
- 3. Physical Design:** Translating the logical design into a physical layout on a silicon wafer, determining how components will be placed and interconnected.
- 4. Verification and Testing:** Ensuring the chip meets all functional and physical requirements through simulations and testing.

Modern chip design is influenced by the need to maximize performance while minimizing power consumption, size, and manufacturing costs. Constraints such as heat dissipation, power routing, and interconnect delay also make the design process highly complex and computationally intensive

## 1.2 The Goal of This Project

- 1. Maximizing Space Utilization:**
  - Minimize wasted space to allow for smaller, cost-effective chip designs.
- 2. Adhering to Constraints:**
  - Manage constraints like power consumption, thermal dissipation, and adjacency rules to ensure safe and efficient chip operation.
- 3. Minimizing Gaps and Wastage:**
  - Reduce unused spaces to improve efficiency and cut manufacturing costs.
- 4. Improving Adaptability:**
  - Create solutions that are effective across various chip sizes and layouts.
- 5. Algorithmic Approaches:**

- Use greedy placement, heuristic methods, and grid-based optimization to balance simplicity, efficiency, and precision.

#### **6. Applications Beyond Chip Design:**

- Techniques also applicable in logistics and other domains requiring efficient packing solutions.

## CHAPTER 2 : PROBLEM STATEMENT

### 2.1 Statement:

This project aims to develop and evaluate algorithms for component placement in chip design. The focus is on creating methods that maximize precision and efficiency while considering practical constraints like power, adjacency, and thermal dissipation.

### 2.2 Common Programming Problems

Let's focus on some key programming challenges in chip design:

1. **Component Placement:** This involves arranging various-sized blocks, like transistors or memory units, on a chip without overlaps. It's like solving a puzzle where every piece must fit perfectly.
2. **Space Utilization:** Here, the goal is to maximize the use of available area while avoiding gaps. Efficient space usage directly impacts chip size and cost.
3. **Order of Placement:** Finding the optimal sequence for arranging components is crucial to minimize delays and ensure smooth manufacturing.
4. **Flexibility Constraints:** Some components have fixed dimensions or orientations, adding to the complexity.

These challenges are inspired by classic computational problems like the bin-packing problem or the knapsack problem, but with added real-world constraints that make them even more interesting.

### 2.3 Importance of Solving These Problems

Why is solving these problems so important?

- **Cost Reduction:** Efficient designs minimize material waste and manufacturing costs.
- **Performance Boost:** Optimized layouts improve data transmission and chip speed.
- **Scalability:** As technology evolves, we need to fit more functionality into the same space.
- **Environmental Impact:** Efficient designs also support sustainable production by reducing resource usage.

### 2.4 Specific Problem Addressed

In my project, I tackled the problem of efficient placement of components within a chip. The task was to arrange smaller rectangular blocks inside a larger rectangular area, ensuring:

1. **No Overlaps:** Each block is placed without overlapping another.

2. **Maximized Space Usage:** Optimal utilization of the available area.
3. **Power Constraints:** Addressed scenarios where adjacent blocks generate heat or power above a set threshold, ensuring such blocks are not placed next to each other.

To achieve this, I implemented:

- **Exploration of Configurations:** Tried various layouts to identify the most efficient arrangement.
- **Visualization Tools:** Developed visual aids to understand and refine the placement strategy.
- **Efficiency Metrics:** Calculated area utilization to measure and improve the results.
- **Advanced Strategies:**
  - Filling gaps to avoid wasted space.
  - Starting placements from varied positions to explore diverse possibilities.

This comprehensive approach ensures optimal performance while adhering to practical constraints like power and thermal management.

## CHAPTER 3 : PROPOSED SOLUTIONS AND WORK DONE

### 3.1 Methods and Tools Used:

To address the problem of component placement in chip design, I employed the following methods and tools:

#### 1. Methods:

- Inspired by the bin-packing problem, the project aimed to optimize space utilization by testing multiple placement orders and configurations.
- A **greedy heuristic algorithm** was applied to fit components sequentially, ensuring efficient usage of space.
- A **grid-based optimization approach** was developed for higher precision and adherence to constraints.

#### 2. Tools:

- **Programming Language:** Python
- **Visualization:** Matplotlib for plotting and visual representation of placements.
- **Data Handling:** File handling techniques were used for reading input dimensions and constraints.

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### 3.2 Step-by-Step Process (Input Stage)

#### Overview

The process begins by reading essential input data from a text file. This data includes:

1. **Dimensions of the large box:** Specifies the total available space.
2. **Dimensions of smaller boxes:** Details the size of each component to be placed.
3. **Power consumption of each component:** Adds a realistic constraint, simulating real-world chip design challenges.

```
10 10
1 5 1
7 2 2
2 2 1
1 1 1
5 5 4
5 5 4
4 4 3
3 4 2
3 2 2
2 2 5
```



## Key Features

- **Power Thresholds:** Ensures adjacent blocks remain within safe power limits.
- **Implementation:** File I/O and validation for accurate inputs.

## 3.3 Algorithmic Logic

### Core Conditions

The placement logic relies on two key checks for each block:

1. **Fit Test:** The block must fit into the available space without overlapping with existing blocks.
2. **Power Constraint Test:** Adjacent blocks must comply with the power threshold.

### Optimization Techniques

- **Placement Sequences:** Various orders of placing the blocks are tested to determine the configuration that maximizes area utilization.
- **Gap Filling:** After filling rows or columns, smaller gaps are revisited to place remaining blocks, maximizing efficiency.
- **Alternative Starting Positions:**
  - Placement starts from different corners or strategic positions within the large box.
  - Configurations are explored systematically to identify the best arrangement.

### Metrics for Success

- **Area Utilization:** The percentage of available space effectively used.
  - **Component Fit:** A record of which blocks were successfully placed and which were not.
- 

## 3.4 Output Visualization and Metrics

### Key Metrics

1. **Total Area Utilized vs. Available Area:** A quantitative measure of how efficiently the space is used.
2. **Percentage Utilization:** Highlights the effectiveness of the packing algorithm.
3. **Component Status:** Clearly indicates:

- Successfully placed blocks.
- Unplaced blocks and reasons for exclusion (e.g., size, power constraint).

### Visual Representation

- **Labeled Plots:**
  - Each block is displayed with its dimensions and power consumption.
  - Exceeds the threshold are visually marked to emphasize violations.
- **Comparative Analysis:** Multiple configurations are visualized to illustrate trade-offs and improvements.

### Benefits of Visualization

- Provides clear insights into the arrangement.
- Helps identify areas for improvement in the algorithm.
- Facilitates comparison between configurations for better decision-making.

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## 3.5 Introduction to the First Algorithm

The first step in this journey towards space optimization was designing an algorithm to efficiently pack smaller boxes into a larger box. The primary aim of this algorithm is to maximize space utilization without any overlaps.

### Core Process:

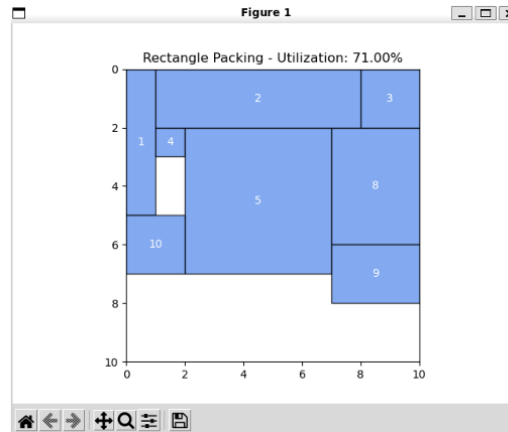
- **Input:** The algorithm takes the dimensions of the large box and smaller boxes as input.
- **Greedy Placement Approach:**
  - Boxes are placed sequentially, starting from the top-left corner.
  - Each box is checked for fit in the remaining space.
  - As many boxes as possible are placed, maximizing the area usage.

### Key Results:

For instance, with a 10x10 large box and smaller boxes of varying dimensions, the algorithm achieved **71% utilization**. While this is a significant result, certain limitations like unplaced boxes and leftover gaps reduced efficiency.

### Real-World Applications:

This algorithm is relevant in fields like logistics, manufacturing, and layout design, where efficient packing is critical for cost and resource optimization.



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### 3.6 Second Algorithm introducing Power Constraint

Building upon the first algorithm, the second version introduces a **power constraint** to reflect real-world considerations in chip design.

#### Objective:

To arrange small boxes within a large box while maximizing space utilization and ensuring that adjacent boxes do not exceed a predefined power threshold.

#### Key Features:

- **Input:**
  - Dimensions of the large and small boxes.
  - Power consumption values for each small box.
- **Enhanced Logic:**
  - Boxes are placed in different orders to find the configuration with the highest space utilization.
  - Adjacent boxes are monitored for their combined power consumption, ensuring the threshold is not exceeded.

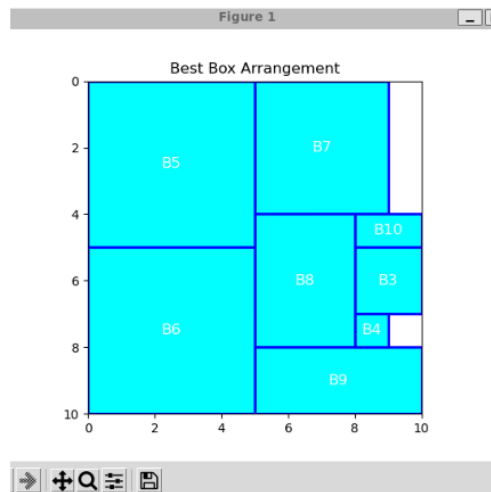
#### Results:

With a 10x10 large box and small boxes having specific dimensions and power values, the algorithm achieved **93% utilization**.

#### Applications:

This method is particularly relevant in:

- **Electronics:** Chip layout design, where power dissipation is critical.
- **Packaging:** Ensuring safe placement of sensitive materials.



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### 3.7 Heuristic-Based Approach

To improve upon the previous methods, the third algorithm employs a **heuristic-based approach** for better efficiency and adaptability.

#### Process Overview:

1. **Sorting:** Boxes are sorted by area in descending order, prioritizing larger boxes to minimize gaps.
2. **Placement Logic:**
  - Iteratively checks potential positions for each box in the large box.
  - Considers both original and rotated orientations to maximize fit.
3. **Greedy Approach:** Ensures placement of as many boxes as possible without overlaps.

#### Key Results:

Using a 10x10 large box and smaller boxes, this approach achieved **90% area utilization**.

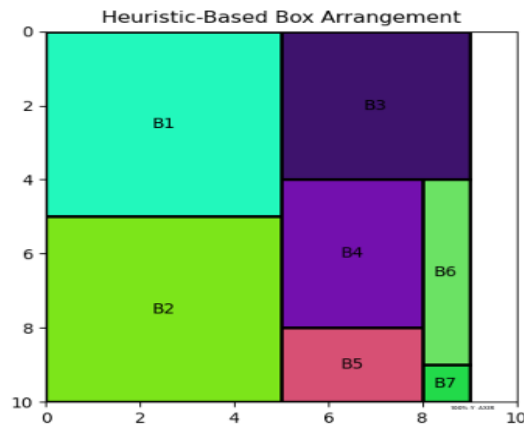
#### Advantages of the Heuristic-Based Approach:

- Simpler and faster for less complex scenarios.
- Effective in handling a variety of box dimensions.
- Minimizes gaps compared to the greedy placement approach.

#### Applications:

This method is ideal for:

- Storage optimization.
- Layout design where rapid calculations are needed.




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### 3.8 Grid-Based Optimization

The fourth version of the algorithm introduces a **grid-based optimization method**, which significantly enhances efficiency and precision.

#### How It Works:

1. **Grid Division:** The large box is divided into a fine-grained grid, where each cell represents a potential placement location.
2. **Placement Evaluation:**
  - Each box is checked against the grid for fit.
  - Both original and rotated orientations are considered.
3. **Visualization:** The grid structure allows for clear representation of the final layout.

#### Results:

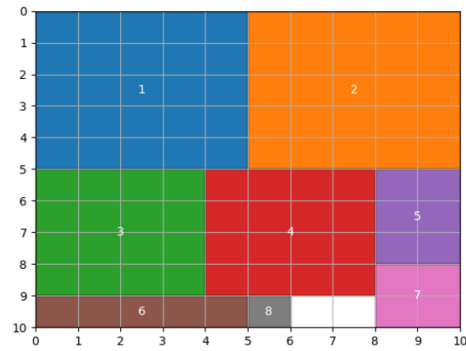
With a 10x10 large box, this approach achieved an impressive **94% utilization**, minimizing gaps and maximizing efficiency.

#### Why the Grid-Based Approach Excels:

- **Optimal Utilization:** Precise placement and alignment minimize wasted space.
- **Constraint Handling:** Effectively manages additional constraints like power thresholds and adjacency rules.
- **Scalability:** Suitable for large-scale or complex problems involving irregular inputs.

#### Applications:

- **Chip Design:** Ensures precise component placement under multiple constraints.
- **Logistics and Storage:** Optimizes space usage in warehouses or transport containers



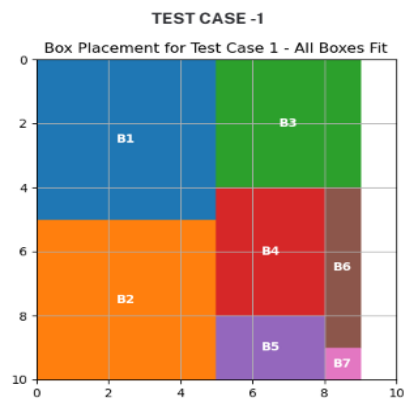
### 3.9 Test Cases and Insights

The final step is evaluating the algorithm across various test cases to measure its adaptability and efficiency.

#### Key Test Cases:

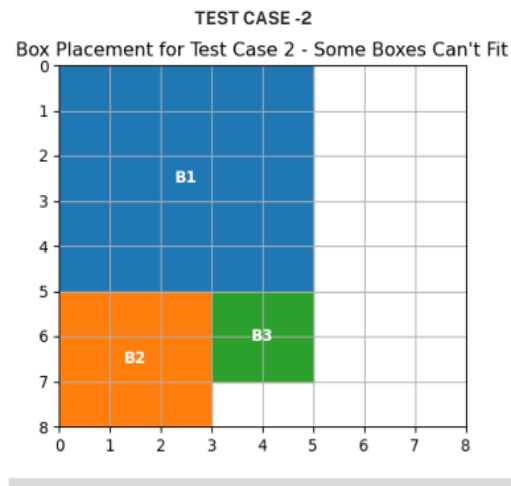
##### 1. Test Case 1:

- **Scenario:** All boxes fit perfectly into a 10x10 grid.
- **Result:** Achieved the highest utilization of **90%**.



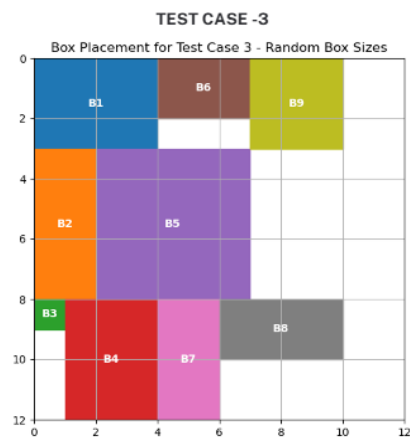
##### 2. Test Case 2:

- **Scenario:** Smaller 8x8 grid.
- **Result:** Some boxes remained unplaced, with utilization dropping to **60%**.

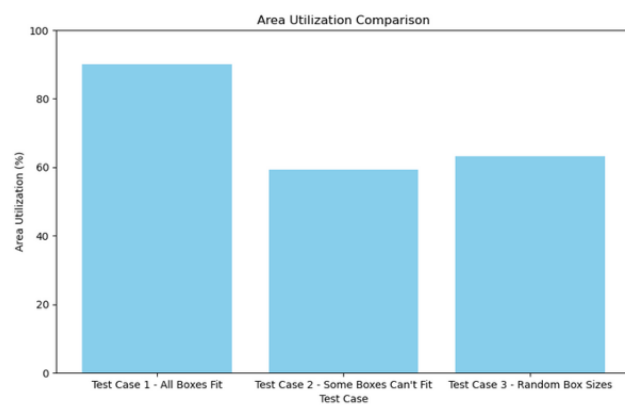


### 3. Test Case 3:

- **Scenario:** Randomly sized boxes in a 12x12 grid.
- **Result:** Achieved **75% utilization**, highlighting challenges with irregular sizes.



The bar chart here compares utilization across test cases, showing the algorithm's strengths and limitations.



## Insights:

- **Strengths:**
  - Efficient packing across diverse scenarios.
  - Visual outputs for easy interpretation.
- **Limitations:**
  - Gaps from irregular sizes.
  - Challenges with overpacking.

## Evaluation Metrics:

- **Area Utilization:** Measures how effectively space is used.
- **Placement Success Rate:** Tracks how many boxes are successfully placed.
- **Visualization:** Highlights areas for improvement.
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## Applications:

This structured evaluation ensures that the algorithm is robust and reliable for real-world applications, such as:

- Storage and logistics.
- Packaging of goods with varying sizes.
- Chip design with complex constraints.



## CHAPTER 4 : RESULTS

### 4.1 Key Findings and Performance

The performance and ideal applications of each algorithm, focusing on their strengths and unique capabilities.

#### 1. Simple and Efficient Packing (First Code):

- **Strengths:** Straightforward and fast. Useful when simplicity is more important than precision.
- **Application:** Suitable for logistics scenarios where the primary focus is to quickly pack boxes with minimal complexity.
- **Example:** Packing items into a delivery van for small-scale operations.

#### 2. Constraint-Driven Applications (Second Code):

- **Strengths:** Handles additional constraints, such as power thresholds or adjacency rules.
- **Application:** Ideal for chip design or packaging scenarios where power dissipation or spatial constraints are critical.
- **Example:** Packing high-power components on a circuit board while ensuring heat dissipation requirements are met.

#### 3. Maximum Precision and Utilization (Grid-Based Code):

- **Strengths:** Delivers the highest utilization by minimizing wasted space. Allows rotations and precise placements.
- **Application:** Complex tasks like designing efficient storage systems or chip layouts for high-performance environments.
- **Example:** Maximizing wafer utilization in silicon fabrication for semiconductors.

#### 4. Comprehensive Algorithm Evaluation (Test Case-Based Approach):

- **Strengths:** Provides robust validation by testing algorithms under real-world constraints like irregular shapes or overpacking.
- **Application:** Useful for benchmarking and refining algorithms in varied scenarios.
- **Example:** Evaluating packing algorithms for diverse manufacturing or logistics problems.

Each algorithm is tailored for specific scenarios, ensuring the best performance based on requirements.

## 4.2 Key Differences Between Algorithms

A detailed comparison of the algorithms based on their objectives, performance, and applications

Feature	First Code (Basic Packing)	Second Code(Power-Constarined Packing)	Heuristic-Based Code	Grid-Bases Optimized Code	Test Case-Based Approach
Primary Objective	Simple packing,maximize maximize utilization	Packing with constraints (Power Consumption)	Prioritize larger boxes for efficiency	Grid-level precision for optimal utilization	Evaluate Algorithm performance undr various scenarios
Flexibility	Limited to simple packing	Handles constraints but fixed placement	Limited flexibility,sorts by size	High flexibility,handles rotations	Converse diverse scenarios (overpacking irregular sizes)
Performance Analysis	No analysis,simple results	Evaluates packing under constraints	Modderate,focused on size-based logic	Precise placement,optimiz ed utilization	Quantifies and compares performance across different scenario's

While the First Code is suitable for basic tasks, the Grid-Based and Heuristic-Based Codes are better for precision and optimization under constraints.

## 4.3 Maximum Utilization vs. More Box Placements

A balanced approach is crucial to address the trade-offs between maximizing functional density and achieving optimal utilization. Focus on Performance, Power, and Area ensures practical and effective solutions for real-world applications.

Aspect	Placing Maximum Boxes	Maximumizing Utilization
Why It's Important	<ul style="list-style-type: none"><li>- High functional density.</li><li>- Supports future workloads.</li><li>- Essential for high-performance chips (e.g., CPUs).</li></ul> Ex: Multi-core processors, GPUs, logic-heavy chips.	<ul style="list-style-type: none"><li>- Efficient area use.</li><li>- Reduces silicon cost.</li><li>- Simplifies interconnections, lowering power.</li></ul> Ex: Memory chips, IoT devices, compact SoCs.
When Prioritized	Fixed chip size, high logic density needed.	Fixed component count, focus on PPA optimization.
Drawbacks	<ul style="list-style-type: none"><li>-Potential gaps or wasted area.</li><li>- Creates thermal hotspots.</li></ul>	<ul style="list-style-type: none"><li>- Risk of overlooking critical components.</li><li>- Potential power density issues.</li></ul>
Key Considerations	Thermal management.	Cost reduction, manufacturability.
Conclusion	Focus on functionality and performance.	Focus on efficiency and cost-effectiveness.
Balanced Approach	Optimize for Performance, Power, and Area (PPA).	

## CHAPTER 5 : FUTURE WORK

### 1. Enhanced Placement Algorithms:

- **Plan:** Incorporate advanced techniques like genetic algorithms, simulated annealing, and machine learning.
- **Objective:** Optimize component placement dynamically, considering constraints like thermal dissipation and power routing.
- **Example:** Using genetic algorithms to explore multiple placement configurations simultaneously for better solutions.

### 2. Interconnect and Routing Optimization:

- **Focus:** Reduce wirelength, minimize delays, and address signal integrity challenges.
- **Objective:** Improve routing efficiency for complex chip designs.
- **Example:** Shorter interconnects in a CPU design to reduce latency and improve clock speeds.

### 3. AI for Layout Optimization:

- **Plan:** Use machine learning models to predict optimal placement strategies based on historical and simulation data.
- **Objective:** Enhance precision and speed in layout optimization.
- **Example:** A neural network that identifies optimal configurations for minimizing thermal hotspots.

### 4. Parallel Processing for Optimization:

- **Plan:** Leverage multi-threaded or distributed algorithms for faster computations.
- **Objective:** Handle large-scale chip designs more efficiently.
- **Example:** Using parallel processing to divide and conquer layout optimization for multi-core processors.

### 5. User-Friendly Interface for Design:

- **Plan:** Develop GUI-based tools for chip designers.
- **Objective:** Simplify the input of constraints and visualization of solutions.
- **Example:** A drag-and-drop interface allowing designers to visualize and tweak placements in real time.

These advancements aim to create efficient, adaptable, and innovative chip design solutions, ensuring future scalability and performance.

## CHAPTER 6 : CONCLUSION

**6.1 Core Focus:** Optimizing space utilization by efficiently arranging components or packing smaller boxes into a larger container.

### 6.2 Key Findings:

- **First Code:** Simple but struggles with complex constraints.
- **Second Code:** Handles power constraints effectively, ideal for energy-efficient scenarios.
- **Heuristic & Grid-Based Codes:** Flexible and achieve high area utilization for complex challenges.
- **Test Case Approach:** Ensures robust performance across varied real-world conditions.

**6.3 Applications:** Relevant in logistics, storage, chip manufacturing, and product design.

**6.4 Benefits:** Leads to cost savings, enhanced performance, and improved manufacturing efficiency.

**6.5 Future Prospects:** Integration of AI, machine learning, and parallel processing for faster and smarter solutions.

**6.6 Vision:** These advancements will drive innovation, enabling smaller, more powerful, and efficient designs.

# CHAPTER 7 : LEARNING FROM BTP

## 7.1 Technical Skills Acquired

- **Algorithm Design:**
  - Developed algorithms for packing problems, focusing on optimization under multiple constraints like space, power, and thermal limits.
  - Gained expertise in heuristic and grid-based approaches for problem-solving.
- **Programming Proficiency:**
  - Improved coding skills in Python, particularly for layout optimization and visualization.
  - Worked with data structures, rotation handling, and grid-based evaluation techniques.
- **Tool Mastery:**
  - Learned to effectively use visualization tools to represent packing solutions and evaluate efficiency metrics.

## 7.2 Analytical and Problem-Solving Skills

- **Logical Thinking:**
  - Addressed complex challenges like irregular box sizes, overlapping constraints, and maximizing utilization while minimizing gaps.
- **Data-Driven Decision Making:**
  - Conducted test case evaluations to identify algorithm strengths and limitations across diverse scenarios.

## 7.3 Practical Insights

- **Real-World Applications:**
  - Understood the significance of packing algorithms in industries like logistics, electronics, and chip manufacturing.
  - Gained awareness of challenges like thermal dissipation and power density in chip designs.
- **Balancing Objectives:**
  - Learned to prioritize between different goals, such as maximizing utilization versus accommodating constraints, to achieve an optimal outcome.

## 7.4 Future Opportunities

- The knowledge gained from this project lays a strong foundation for further exploration in:

- Advanced optimization techniques, including AI and machine learning integration.
- Chip design methodologies focused on reducing energy consumption and improving efficiency.
- Applications in logistics, product design, and layout planning for real-world scenarios.

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