

Project Ideas

1. Pricing Optimization:

Objective:

Develop a pricing model that optimizes revenue or profit while adhering to specific constraints, such as maximum production capacity or minimum acceptable profit margins.

Tasks:

1. Define the Pricing Model:

- Design a pricing model with an objective function (e.g., maximize revenue):

$$\text{Revenue} = \sum_{i=1}^n p_i \cdot d_i(p_i)$$

where p_i is the price and $d_i(p_i)$ is the demand at that price.

- Add constraints like maximum price or demand.
- Solve the optimization problem using **CVXPY**.
 - Use a convex optimization approach to maximize revenue or minimize costs.

2. Determine Convexity:

- Check whether the revenue function is convex using **Hessian matrix** or CVXPY features:
 1. If the function is convex, justify this mathematically.
 2. If not convex, explain why.

3. Modify the Model (Non-Convexity):

- Introduce modifications to *make the model non-convex*:
 1. For example, add a non-linear term in the demand function to make it more complex.

4. Restore Convexity:

- Modify the model again (e.g., remove the non-linear term or simplify the function) to restore convexity.

5. Visualization:

- Use **Matplotlib** or **Seaborn** to plot the relationship between price and revenue in the three cases:
 1. Original (Convex).
 2. Non-Convex.
 3. Restored Convex.

Outcomes:

- Python script implementing the optimization problem.
 - A presentation or report summarizing:
 - The problem setup (objective and constraints).
 - Key insights from the analysis.
 - Visualizations showing pricing vs. revenue trends.
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2. Financial Data Analysis:

Objective:

Analyze a dataset of *financial transactions* or *stock prices* to identify patterns, optimize investment portfolios, or improve financial performance.

Tasks:

1. Use a simple real dataset.

2. Setup a Financial Model:

- Define an objective function (e.g., maximize returns, minimize risk, ...).
- Define the constraints.
- Formulate the Problem.
- Solve the optimization problem using **CVXPY**.
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3. Determine Convexity:

- Check whether the objective function is convex using properties of convex functions (e.g., variance is always convex,) or CVXPY features:
 1. If the function is convex, *justify this mathematically*.
 2. If not convex, explain why.

4. Modify the Model (Non-Convexity):

- Introduce non-linear or varying components to modify the portfolio and make it non-convex:
 1. For example, add non-linear constraints to stock allocations.

5. Restore Convexity:

- Simplify the model (e.g., remove non-linear constraints or use a convex approximation).

6. Visualization:

- Use Matplotlib or Seaborn to visualize financial trends, portfolio performance, or risk-return tradeoffs.
- Use graphs to display the relationship between risk and return in the three cases:
 1. Original (Convex).
 2. Non-Convex.
 3. Restored Convex.

Outcomes:

- Python script implementing the optimization problem.
 - A presentation or report summarizing:
 - The problem setup (objective and constraints).
 - Key findings from the analysis.
 - Visualizations (e.g., portfolio allocation, stock trends).
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3. Students' Exam Scores Analysis:

Objective:

Analyze student exam scores, verify the convexity of the score distribution, modify the distribution to make it non-convex, and then restore convexity.

Tasks:

1. Analyze the Scores and (Optimization):

- Analyze the student score data and its distribution:
 - Example: A simple distribution with mean and standard deviation.
 - **Objective function:** Minimize variance among scores for fairness or the goal could be to improve grades or reduce differences between students' grades (balance students' grades).
 - **Constraints:** Impose some constraints such as minimum grades or maximum grades in each subject.
 - Solve the problem using CVXPY.

2. Determine Convexity:

- Check the convexity of the distribution:
 - Use CVXPY to analyze the objective function.
 - Justify mathematically whether the distribution function is convex or not.

3. Modify the Distribution (Non-Convexity):

- Introduce modifications to the data to make it non-convex:
 - Example: Add outliers or change the grades in a non-linear manner.

4. Restore Convexity:

- Use techniques like:
 - Removing outliers.
 - Adding a constraint to the distribution to make it more organized.

5. Visualization:

- Display the distribution before and after modification using graphs.
- Compare the three cases:
 - Original (Convex).
 - Modified (Non-Convex).
 - Restored (Convex).

Outcomes:

- Python script implementing the analysis and optimization.
 - A presentation or report summarizing:
 - The problem setup (objective and constraints).
 - Key findings from the analysis.
 - Visualizations comparing original and adjusted scores.
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General Guidelines

1. Libraries Required:

- **CVXPY**: For solving optimization problems.
- **Matplotlib/Seaborn**: For visualizing results.

2. Report Structure (or power-point presentation):

- Brief introduction to the problem.
- Description of the dataset and methodology.
- Key insights and recommendations.
- Visualizations with proper labeling and explanations.

3. **In all projects**, each team will **select a real dataset**, then **use CVXPY** to analyze the **constraints** and **make modifications** as described in the requirements.