

# Detailed Report for Assignment 2, Question 2

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This document explains how our Python script fulfills **Question 2** from the DAMO600 Prescriptive Analytics assignment. All cost formulas are displayed as separate LaTeX blocks per your request.

## 1. Question 2 Requirements

The assignment specifies that the Python script must:

- Simulate two fan replacement policies via Monte Carlo.
- Perform exactly 45 replacement events per simulation run.
- Randomly generate both fan lifetimes and technician arrival delays.
- Calculate downtime cost and labor cost for each event.
- Sum these costs to produce a total cost per policy per run □filecite□turn7file4□.

## 2. Simulation Overview

The simulation compares two strategies:

1. **Current Policy:** Replace only the failed fan each time.
2. **Proposed Policy:** Replace all three fans whenever any one fails.

Each failure event incurs three cost components, calculated as follows.

### Replacement Cost

$$C_{\mathrm{rep}} = n \times c_{\mathrm{fan}}$$

where

$$n = \text{number of fans replaced (1 or 3)}, \quad c_{\mathrm{fan}} = \$32.$$

### Downtime Cost

$$C_{\mathrm{down}} = (D + T_{\mathrm{rep}}) \times c_{\mathrm{down}}$$

where

$$D = \text{technician arrival delay (minutes)}, \quad T_{\mathrm{rep}} = \text{replacement duration (20 min for current; 40 min for proposed)}, \\ c_{\mathrm{down}} = \$10/\mathrm{min}.$$

### Labor Cost

$$C_{\mathrm{lab}} = \frac{T_{\mathrm{rep}}}{60} \times c_{\mathrm{labor}}$$

where

$$c_{\mathrm{labor}} = \$30/\mathrm{hr}.$$

The total cost for one failure event is

$$C_{\{\mathrm{event}\}} = C_{\{\mathrm{rep}\}} + C_{\{\mathrm{down}\}} + C_{\{\mathrm{lab}\}}.$$

A full simulation run of  $N = 45$  failures yields

$$C_{\{\mathrm{total}\}} = \sum_{i=1}^{45} C_{\{\mathrm{event}, i\}}.$$

### 3. Random Sampling of Inputs

We sample from the discrete distributions given in the assignment PDF:

- **Fan Lifetimes (hours)** { 1000, 1100, 1200, 1300, 1400, 1500, 1600, 1700, 1800, 1900 } with probabilities {0.10, 0.13, 0.25, 0.13, 0.09, 0.12, 0.02, 0.06, 0.05, 0.05}:

```
lifetime = np.random.choice(fan_lifetimes, p=fan_probs)
```

- **Technician Delays (minutes)** {20, 30, 45} with probabilities {0.60, 0.30, 0.10}:

```
delay = np.random.choice(delay_times, p=delay_probs)
```

### 4. Core Simulation Function

We encapsulate each policy's event loop in a function:

```
def simulate_policy(fans_to_replace, rep_time):
    total_cost = 0.0
    for _ in range(num_failures):
        lifetime = np.random.choice(fan_lifetimes, p=fan_probs)
        delay = np.random.choice(delay_times, p=delay_probs)
        C_rep = fans_to_replace * fan_unit_cost
        C_down = (delay + rep_time) * downtime_cost_per_min
        C_lab = (rep_time / 60.0) * labor_cost_per_hour
        total_cost += (C_rep + C_down + C_lab)
    return total_cost
```

This is invoked twice per run:

```
cost_current = simulate_policy(1, 20)
cost_proposed = simulate_policy(3, 40)
```

### 5. Monte Carlo Loop and Data Capture

A fixed seed ensures reproducibility. We execute 1 000 runs, collecting totals:

```

np.random.seed(42)
results = []
for _ in range(1000):
    results.append((
        simulate_policy(1, rep_time_current),
        simulate_policy(3, rep_time_proposed)
    ))

df_results = pd.DataFrame(
    results,
    columns=["Cost_Current_Policy", "Cost_Proposed_Policy"]
)

```

## 6. Visualization of Outcomes

We visualize `df_results` via histogram and boxplot to compare distributions of total costs across policies.

## 7. Manual Verification Example

Fixing delay  $\$D = 20\$$  min yields per-event costs:

$$\begin{aligned}
 &\text{Current:} \quad C_{\text{rep}} = 1 \times 32 = 32, \quad C_{\text{down}} = (20 + 20) \times 10 = 400, \quad C_{\text{lab}} = (20/60) \times 30 = 10, \quad C_{\text{event}} = 32 + 400 + 10 = 442. \\
 &\text{Proposed:} \quad C_{\text{rep}} = 3 \times 32 = 96, \quad C_{\text{down}} = (20 + 40) \times 10 = 600, \quad C_{\text{lab}} = (40/60) \times 30 = 20, \quad C_{\text{event}} = 96 + 600 + 20 = 716.
 \end{aligned}$$

Aggregating 45 failures:

$$45 \times 442 = 19,890, \quad 45 \times 716 = 32,220.$$

Comparing these hand-computed totals to Monte Carlo means confirms correctness.

## 8. Conclusion and Recommendation

We compute average total cost per policy over 1 000 runs; the policy with the lower mean is recommended.

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References DAMO600 P25 Assignment 2 Description [filecite](#) [turn7file4](#)