## experimento 6: gemini - nao rodou - 2d96f53eb

## **USER**

A base de dados anexada no arquivo dados\_50\_50.csv possui quatro variáveis: IDENTIFICADOR, CARGO, PERCENT\_PRI E DISTANCE\_KM. A primeira variável é o codigo que identifica o funcionário, a segunda apresenta os códigos dos cargos, a terceira é um indice que varia de 0 a 1, sendo o 0 o melhor e, por fim, a última variável é a distância em KM entre o funcionário e o cargo que ele poderá assumir, ou seja, quanto maior a distância, pior. Gostaria que me propusesse uma metodologia para solucionar meu problema de alocação de forma a minimizar o PERCENT\_PRI ou a DISTANCE\_KM ou os dois juntos. Determine e execute o código para solucionar o problema e indique o custo total minimizado.

## BOT

Here's a proposed methodology and solution for your allocation problem:

\*\*1. Problem Definition:\*\*

You're dealing with an assignment problem. You want to assign each employee (IDENTIFICADOR) to a single job (CARGO) in a way that minimizes either PERCENT PRI, DISTANCE KM, or a combination of both.

- \*\*2. Methodology:\*\*
- \* \*\*Separate Optimization:\*\* If you want to minimize only one of the objectives (PERCENT\_PRI or DISTANCE\_KM), the Hungarian algorithm is a good choice. It's efficient for solving assignment problems with a single objective.
- \* \*\*Combined Optimization:\*\* If you want to minimize both, you have a multi-objective optimization problem. There isn't a single "best" solution in this case, but rather a set of Pareto optimal solutions. A Pareto optimal solution is one where you can't improve one objective without worsening the other. One approach is to create a weighted sum objective function. This involves assigning weights to each objective and summing them. The weights reflect the relative importance of each objective.
- \*\*3. Chosen Method and Implementation (Weighted Sum):\*\*

Given that you mentioned minimizing "one or both," I'll implement a weighted sum approach, allowing you to tune the weights. This provides flexibility.

```
```python
import pandas as pd
from scipy.optimize import linear_sum_assignment
# Load the data
df = pd.read_csv("dados_50_50.csv", sep=";")
# Convert ',' to '.' and to numeric
df['PERCENT PRI'] = pd.to numeric(df['PERCENT PRI'].str.replace(',', ','))
df['DISTANCE KM'] = pd.to numeric(df['DISTANCE KM'].str.replace(',', '.'))
# Define weights for the objectives
weight percent pri = 0.5 # Example: Equal weight
weight distance km = 0.5
# Create the cost matrix
cost matrix = df.pivot(index='IDENTIFICADOR', columns='CARGO', values=['PERCENT PRI', 'DISTANCE KM'])
# Normalize the objectives (important for weighted sum)
cost_matrix = cost_matrix.apply(lambda x: (x - x.min()) / (x.max() - x.min()))
# Calculate the weighted sum cost
cost matrix['weighted cost'] = weight percent pri * cost matrix[('PERCENT PRI', )] + weight distance km *
cost_matrix[('DISTANCE_KM', )]
cost_matrix = cost_matrix['weighted_cost'].values
# Apply the Hungarian algorithm
```

```
row_ind, col_ind = linear_sum_assignment(cost_matrix)

# Calculate the total minimized cost
total_cost = cost_matrix[row_ind, col_ind].sum()

# Print the results
print(f"Total minimized weighted cost: {total_cost}")

# Map the solution back to the original data
solution = pd.DataFrame({'IDENTIFICADOR': row_ind + 1, 'CARGO': col_ind + 1}) # +1 because indices start from 0
print("\nOptimal Assignment:")
print(solution)

# Created/Modified files during execution:
# No files created/modified

***4. Explanation and Next Steps:**
```

The code first loads and preprocesses the data, converting the string representations of numbers to actual numeric values. It then defines weights for each objective. A cost matrix is created and normalized (crucial for the weighted sum to work correctly). The Hungarian algorithm is then used to find the optimal assignment. Finally, the total minimized cost and the assignment are printed.

You can adjust `weight\_percent\_pri` and `weight\_distance\_km` to change the importance of each objective. For example, setting `weight\_percent\_pri` to 1 and `weight\_distance\_km` to 0 would minimize only PERCENT\_PRI. Experiment with different weights to find a solution that best suits your needs. If you have additional constraints or requirements, let me know, and I can refine the code.

```
Execution Error: ValueError
   Traceback (most recent call last)
/tmp/ipykernel_1655/375967982.py in ?()
     18 # Normalize the objectives (important for weighted sum)
     19 cost_matrix = cost_matrix.apply(lambda x: (x - x.min()) / (x.max() - x.min()))
     21 # Calculate the weighted sum cost
  ---> 22 cost matrix['weighted cost'] = weight percent pri * cost matrix[('PERCENT PRI', )] + weight distance km *
cost matrix[('DISTANCE KM', )]
     23 cost_matrix = cost_matrix['weighted_cost'].values
     24
     25 # Apply the Hungarian algorithm
/opt/venv/lib/python3.12/site-packages/pandas/core/frame.py in ?(self, key, value)
    4297
                 self._setitem_frame(key, value)
    4298
               elif isinstance(key, (Series, np.ndarray, list, Index)):
    4299
                 self. setitem array(key, value)
    4300
               elif isinstance(value, DataFrame):
  -> 4301
                  self._set_item_frame_value(key, value)
    4302
               elif (
    4303
                 is list like(value)
    4304
                 and not self.columns.is unique
/opt/venv/lib/python3.12/site-packages/pandas/core/frame.py in ?(self, key, value)
    4456
                 return self.isetitem(locs, value)
    4457
    4458
               if len(value.columns) > 1:
  -> 4459
                  raise ValueError(
    4460
                    "Cannot set a DataFrame with multiple columns to the single "
    4461
                    f"column {key}"
    4462
```

ValueError: Cannot set a DataFrame with multiple columns to the single column weighted\_cost