

Assignment 4
Of
Modelling & Simulation Lab (CS1052)

Masters of Technology in Computer Science And Engineering

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Problem statement 1

Find the optimal solution to the following transportation problem in which the cells contains the unit transportation cost in rupees.

	w_1	w_2	w_3	w_4	w_5	avail.
F_1	7	6	4	5	9	40
F_2	8	5	6	7	8	30
F_3	6	8	9	6	5	20
F_4	5	7	7	8	6	10
Req.	30	30	15	20	5	

use NWCR and LCM for initial basic feasible solution.

Problem formulation -

	w_1	w_2	w_3	w_4	w_5	Supply
I_1	x_{11} 7	x_{12} 6	x_{13} 4	x_{14} 5	x_{15} 9	40
I_2	x_{21} 8	x_{22} 5	x_{23} 6	x_{24} 7	x_{25} 8	30
I_3	x_{31} 6	x_{32} 8	x_{33} 9	x_{34} 6	x_{35} 5	20
Dem.	x_{41} 5	x_{42} 7	x_{43} 7	x_{44} 8	x_{45} 6	10
Dem.	30	30	15	20	5	

The transportation problem is formulated as an LP model as follows -

Minimize (Total T.P cost) -

$$\begin{aligned}
 & 7x_{11} + 6x_{12} + 4x_{13} + 5x_{14} + 9x_{15} \\
 & + 8x_{21} + 5x_{22} + 6x_{23} + 7x_{24} + 8x_{25} \\
 & + 6x_{31} + 8x_{32} + 9x_{33} + 6x_{34} + 5x_{35} \\
 & + 5x_{41} + 7x_{42} + 7x_{43} + 8x_{44} + 6x_{45}
 \end{aligned}$$

Subject to constraint -

$$\begin{aligned}
 & 7x_{11} + 6x_{12} + 4x_{13} + 5x_{14} + 9x_{15} = 40 \\
 & 8x_{21} + 5x_{22} + 6x_{23} + 7x_{24} + 8x_{25} = 30 \\
 & 6x_{31} + 8x_{32} + 9x_{33} + 6x_{34} + 5x_{35} = 20 \\
 & 5x_{41} + 7x_{42} + 7x_{43} + 8x_{44} + 6x_{45} = 10
 \end{aligned}$$

First we will solve it using north west corner rule.

The allocation matrix is such that.

	w_1	w_2	w_3	w_4	w_5	exp
F_1	7 (20)	6 (10)	7	5	9	40 10
F_2	8	5 (20)	6 (10)	7	8	30 10
F_3	6	8	9 (5)	6 (15)	5	20 10
F_4	5	7	7	8 (5)	6	10 0
Demand	30 0	30 20 0	15 50	20 150	50	

The step by step description of this solution is given as :-

and

$$2x_{11} + 8x_{21} + 6x_{31} + 5x_{41} = 30$$

$$6x_{12} + 5x_{22} + 8x_{32} + 7x_{42} = 30$$

$$4x_{13} + 6x_{23} + 9x_{33} + 7x_{43} = 15$$

$$5x_{14} + 7x_{24} + 6x_{34} + 8x_{44} = 20$$

$$9x_{15} + 8x_{25} + 5x_{35} + 6x_{45} = 5$$

and $x_{ij} \geq 0$ for $i = 1, 2, 3, 4$

$j = 1, 2, 3, 4$

Soln

Total no of supply const. = 4

Total no of demand const. = 5

They sum up to 40 & 30 compared.

$$\text{Total supply} = 40 + 30 + 20 + 10 = 100$$

$$\text{Total demand} = 30 + 30 + 15 + 20 + 5 = 100$$

Total supply = Total demand
 \Rightarrow Balanced Transp. problem.

North west corner rule

The sum val $F_1 \geq 40$ & $w_1 \geq 30$
 $\min(F_1, w_1) \geq 30$ is assigned
to F_1, w_1 , this meets the
demand of w_1 & leaves $40 - 30 =$
10 with F_1 .

The sum value for $F_2 \geq 30$ &
 $w_2 \geq 20$ are compared. The
 $\min(F_2, w_2) = 20$. This meets
the demand of w_2 & leaves
 $30 - 20 = 10$ units with F_2 .

The sum values for $F_2 \geq 10$
and $w_3 \geq 15$ are compared.
The smaller of the two
i.e. $\min(10, 15) = 10$ is assign.
to F_2, w_3 , this exhaust

Python Code:

```
class HungarianError(Exception):
    pass

try:
    import numpy as np
except ImportError:
    raise HungarianError("NumPy is not installed.")

class Hungarian:
    """
    Implementation of the Hungarian (Munkres) Algorithm using np.
    Usage:
        hungarian = Hungarian(cost_matrix)
        hungarian.calculate()
    or
        hungarian = Hungarian()
        hungarian.calculate(cost_matrix)
    Handle Profit matrix:
        hungarian = Hungarian(profit_matrix, is_profit_matrix=True)
    or
        cost_matrix = Hungarian.make_cost_matrix(profit_matrix)
    The matrix will be automatically padded if it is not square.
    For that numpy's resize function is used, which automatically adds 0's to any row/c
    Get results and total potential after calculation:
        hungarian.get_results()
        hungarian.get_total_potential()
    """

    def __init__(self, input_matrix=None, is_profit_matrix=False):
        """
        input_matrix is a List of Lists.
        input_matrix is assumed to be a cost matrix unless is_profit_matrix is True.
        """
        if input_matrix is not None:
            # Save input
            my_matrix = np.array(input_matrix)
            self._input_matrix = np.array(input_matrix)
            self._maxColumn = my_matrix.shape[1]
            self._maxRow = my_matrix.shape[0]

            # Adds 0s if any columns/rows are added. Otherwise stays unaltered
            matrix_size = max(self._maxColumn, self._maxRow)
            pad_columns = matrix_size - self._maxRow
            pad_rows = matrix_size - self._maxColumn
            my_matrix = np.pad(my_matrix, ((0, pad_columns), (0, pad_rows)), 'constant')

            # Convert matrix to profit matrix if necessary
            if is_profit_matrix:
                my_matrix = self.make_cost_matrix(my_matrix)
```

```

        self._cost_matrix = my_matrix
        self._size = len(my_matrix)
        self._shape = my_matrix.shape

        # Results from algorithm.
        self._results = []
        self._totalPotential = 0
    else:
        self._cost_matrix = None

def get_results(self):
    """Get results after calculation."""
    return self._results

def get_total_potential(self):
    """Returns expected value after calculation."""
    return self._totalPotential

def calculate(self, input_matrix=None, is_profit_matrix=False):
    """
    Implementation of the Hungarian (Munkres) Algorithm.
    input_matrix is a List of Lists.
    input_matrix is assumed to be a cost matrix unless is_profit_matrix is True.
    """
    # Handle invalid and new matrix inputs.
    if input_matrix is None and self._cost_matrix is None:
        raise HungarianError("Invalid input")
    elif input_matrix is not None:
        self._init__(input_matrix, is_profit_matrix)

    result_matrix = self._cost_matrix.copy()

    # Step 1: Subtract row mins from each row.
    for index, row in enumerate(result_matrix):
        result_matrix[index] -= row.min()

    # Step 2: Subtract column mins from each column.
    for index, column in enumerate(result_matrix.T):
        result_matrix[:, index] -= column.min()

    # Step 3: Use minimum number of lines to cover all zeros in the matrix.
    # If the total covered rows+columns is not equal to the matrix size then adjust
    total_covered = 0
    while total_covered < self._size:
        # Find minimum number of lines to cover all zeros in the matrix and find to
        cover_zeros = CoverZeros(result_matrix)
        covered_rows = cover_zeros.get_covered_rows()
        covered_columns = cover_zeros.get_covered_columns()
        total_covered = len(covered_rows) + len(covered_columns)

    # if the total covered rows+columns is not equal to the matrix size then ad
    if total_covered < self._size:
        result_matrix = self._adjust_matrix_by_min_uncovered_num(result_matrix,

```



```

# Step 4: Starting with the top row, work your way downwards as you make assignments
# Find single zeros in rows or columns.
# Add them to final result and remove them and their associated row/column from
expected_results = min(self._maxColumn, self._maxRow)
zero_locations = (result_matrix == 0)
while len(self._results) != expected_results:

    # If number of zeros in the matrix is zero before finding all the results then
    if not zero_locations.any():
        raise HungarianError("Unable to find results. Algorithm has failed.")

    # Find results and mark rows and columns for deletion
    matched_rows, matched_columns = self._find_matches(zero_locations)

    # Make arbitrary selection
    total_matched = len(matched_rows) + len(matched_columns)
    if total_matched == 0:
        matched_rows, matched_columns = self.select_arbitrary_match(zero_locations)

    # Delete rows and columns
    for row in matched_rows:
        zero_locations[row] = False
    for column in matched_columns:
        zero_locations[:, column] = False

    # Save Results
    self._set_results(zip(matched_rows, matched_columns))

# Calculate total potential
value = 0
for row, column in self._results:
    value += self._input_matrix[row, column]
self._totalPotential = value

@staticmethod
def make_cost_matrix(profit_matrix):
    """
    Converts a profit matrix into a cost matrix.
    Expects NumPy objects as input.
    """
    # subtract profit matrix from a matrix made of the max value of the profit matrix
    matrix_shape = profit_matrix.shape
    offset_matrix = np.ones(matrix_shape, dtype=int) * profit_matrix.max()
    cost_matrix = offset_matrix - profit_matrix
    return cost_matrix

def _adjust_matrix_by_min_uncovered_num(self, result_matrix, covered_rows, covered_columns):
    """ Subtract m from every uncovered number and add m to every element covered with m
    # Calculate minimum uncovered number (m)
    elements = []
    for row_index, row in enumerate(result_matrix):
        if row_index not in covered_rows:
            for index, element in enumerate(row):
                if index not in covered_columns:

```

```

        elements.append(element)
min_uncovered_num = min(elements)

# Add m to every covered element
adjusted_matrix = result_matrix
for row in covered_rows:
    adjusted_matrix[row] += min_uncovered_num
for column in covered_columns:
    adjusted_matrix[:, column] += min_uncovered_num

# Subtract m from every element
m_matrix = np.ones(self._shape, dtype=int) * min_uncovered_num
adjusted_matrix -= m_matrix

return adjusted_matrix

def __find_matches(self, zero_locations):
    """Returns rows and columns with matches in them."""
    marked_rows = np.array([], dtype=int)
    marked_columns = np.array([], dtype=int)

    # Mark rows and columns with matches
    # Iterate over rows
    for index, row in enumerate(zero_locations):
        row_index = np.array([index])
        if np.sum(row) == 1:
            column_index, = np.where(row)
            marked_rows, marked_columns = self.__mark_rows_and_columns(marked_rows,
                                                                       column_index)

    # Iterate over columns
    for index, column in enumerate(zero_locations.T):
        column_index = np.array([index])
        if np.sum(column) == 1:
            row_index, = np.where(column)
            marked_rows, marked_columns = self.__mark_rows_and_columns(marked_rows,
                                                                       column_index)

    return marked_rows, marked_columns

    @staticmethod
def __mark_rows_and_columns(marked_rows, marked_columns, row_index, column_index):
    """Check if column or row is marked. If not marked then mark it."""
    new_marked_rows = marked_rows
    new_marked_columns = marked_columns
    if not (marked_rows == row_index).any() and not (marked_columns == column_index):
        new_marked_rows = np.insert(marked_rows, len(marked_rows), row_index)
        new_marked_columns = np.insert(marked_columns, len(marked_columns), column_index)
    return new_marked_rows, new_marked_columns

    @staticmethod
def select_arbitrary_match(zero_locations):
    """Selects row column combination with minimum number of zeros in it."""
    # Count number of zeros in row and column combinations

```

```

rows, columns = np.where(zero_locations)
zero_count = []
for index, row in enumerate(rows):
    total_zeros = np.sum(zero_locations[row]) + np.sum(zero_locations[:, columns[index]])
    zero_count.append(total_zeros)

# Get the row column combination with the minimum number of zeros.
indices = zero_count.index(min(zero_count))
row = np.array([rows[indices]])
column = np.array([columns[indices]])

return row, column

def __set_results(self, result_lists):
    """Set results during calculation."""
    # Check if results values are out of bound from input matrix (because of matrix
    # Add results to results list.
    for result in result_lists:
        row, column = result
        if row < self._maxRow and column < self._maxColumn:
            new_result = (int(row), int(column))
            self._results.append(new_result)

class CoverZeros:
    """
    Use minimum number of lines to cover all zeros in the matrix.
    Algorithm based on: http://weber.ucsd.edu/~vcrawfor/hungar.pdf
    """

    def __init__(self, matrix):
        """
        Input a matrix and save it as a boolean matrix to designate zero locations.
        Run calculation procedure to generate results.
        """
        # Find zeros in matrix
        self._zero_locations = (matrix == 0)
        self._shape = matrix.shape

        # Choices starts without any choices made.
        self._choices = np.zeros(self._shape, dtype=bool)

        self._marked_rows = []
        self._marked_columns = []

        # marks rows and columns
        self._calculate()

        # Draw lines through all unmarked rows and all marked columns.
        self._covered_rows = list(set(range(self._shape[0])) - set(self._marked_rows))
        self._covered_columns = self._marked_columns

    def get_covered_rows(self):
        """Return list of covered rows."""

```

```

        return self._covered_rows

def get_covered_columns(self):
    """Return list of covered columns."""
    return self._covered_columns

def __calculate(self):
    """
    Calculates minimum number of lines necessary to cover all zeros in a matrix.
    Algorithm based on: http://weber.ucsd.edu/~vcrawfor/hungar.pdf
    """
    while True:
        # Erase all marks.
        self._marked_rows = []
        self._marked_columns = []

        # Mark all rows in which no choice has been made.
        for index, row in enumerate(self._choices):
            if not row.any():
                self._marked_rows.append(index)

        # If no marked rows then finish.
        if not self._marked_rows:
            return True

        # Mark all columns not already marked which have zeros in marked rows.
        num_marked_columns = self.__mark_new_columns_with_zeros_in_marked_rows()

        # If no new marked columns then finish.
        if num_marked_columns == 0:
            return True

        # While there is some choice in every marked column.
        while self.__choice_in_all_marked_columns():
            # Some Choice in every marked column.

            # Mark all rows not already marked which have choices in marked columns
            num_marked_rows = self.__mark_new_rows_with_choices_in_marked_columns()

            # If no new marks then Finish.
            if num_marked_rows == 0:
                return True

            # Mark all columns not already marked which have zeros in marked rows.
            num_marked_columns = self.__mark_new_columns_with_zeros_in_marked_rows()

            # If no new marked columns then finish.
            if num_marked_columns == 0:
                return True

        # No choice in one or more marked columns.
        # Find a marked column that does not have a choice.
        choice_column_index = self.__find_marked_column_without_choice()

```



```

while choice_column_index is not None:
    # Find a zero in the column indexed that does not have a row with a cho
    choice_row_index = self._find_row_without_choice(choice_column_index)

    # Check if an available row was found.
    new_choice_column_index = None
    if choice_row_index is None:
        # Find a good row to accomodate swap. Find its column pair.
        choice_row_index, new_choice_column_index = self._find_best_choice_r
        choice_column_index)

    # Delete old choice.
    self._choices[choice_row_index, new_choice_column_index] = False

    # Set zero to choice.
    self._choices[choice_row_index, choice_column_index] = True

    # Loop again if choice is added to a row with a choice already in it.
    choice_column_index = new_choice_column_index

def _mark_new_columns_with_zeros_in_marked_rows(self):
    """Mark all columns not already marked which have zeros in marked rows."""
    num_marked_columns = 0
    for index, column in enumerate(self._zero_locations.T):
        if index not in self._marked_columns:
            if column.any():
                row_indices, = np.where(column)
                zeros_in_marked_rows = (set(self._marked_rows) & set(row_indices))
                if zeros_in_marked_rows:
                    self._marked_columns.append(index)
                    num_marked_columns += 1
    return num_marked_columns

def _mark_new_rows_with_choices_in_marked_columns(self):
    """Mark all rows not already marked which have choices in marked columns."""
    num_marked_rows = 0
    for index, row in enumerate(self._choices):
        if index not in self._marked_rows:
            if row.any():
                column_index, = np.where(row)
                if column_index in self._marked_columns:
                    self._marked_rows.append(index)
                    num_marked_rows += 1
    return num_marked_rows

def _choice_in_all_marked_columns(self):
    """Return Boolean True if there is a choice in all marked columns. Returns bool
    for column_index in self._marked_columns:
        if not self._choices[:, column_index].any():
            return False
    return True

def _find_marked_column_without_choice(self):
    """Find a marked column that does not have a choice."""

```

```

        for column_index in self._marked_columns:
            if not self._choices[:, column_index].any():
                return column_index

        raise HungarianError(
            "Could not find a column without a choice. Failed to cover matrix zeros. All

def __find_row_without_choice(self, choice_column_index):
    """Find a row without a choice in it for the column indexed. If a row does not
    row_indices, = np.where(self._zero_locations[:, choice_column_index])
    for row_index in row_indices:
        if not self._choices[row_index].any():
            return row_index

    # All rows have choices. Return None.
    return None

def __find_best_choice_row_and_new_column(self, choice_column_index):
    """
    Find a row index to use for the choice so that the column that needs to be changed.
    Return a random row and column if unable to find an optimal selection.
    """
    row_indices, = np.where(self._zero_locations[:, choice_column_index])
    for row_index in row_indices:
        column_indices, = np.where(self._choices[row_index])
        column_index = column_indices[0]
        if self.__find_row_without_choice(column_index) is not None:
            return row_index, column_index

    # Cannot find optimal row and column. Return a random row and column.
    from random import shuffle

    shuffle(row_indices)
    column_index, = np.where(self._choices[row_indices[0]])
    return row_indices[0], column_index[0]

if __name__ == '__main__':

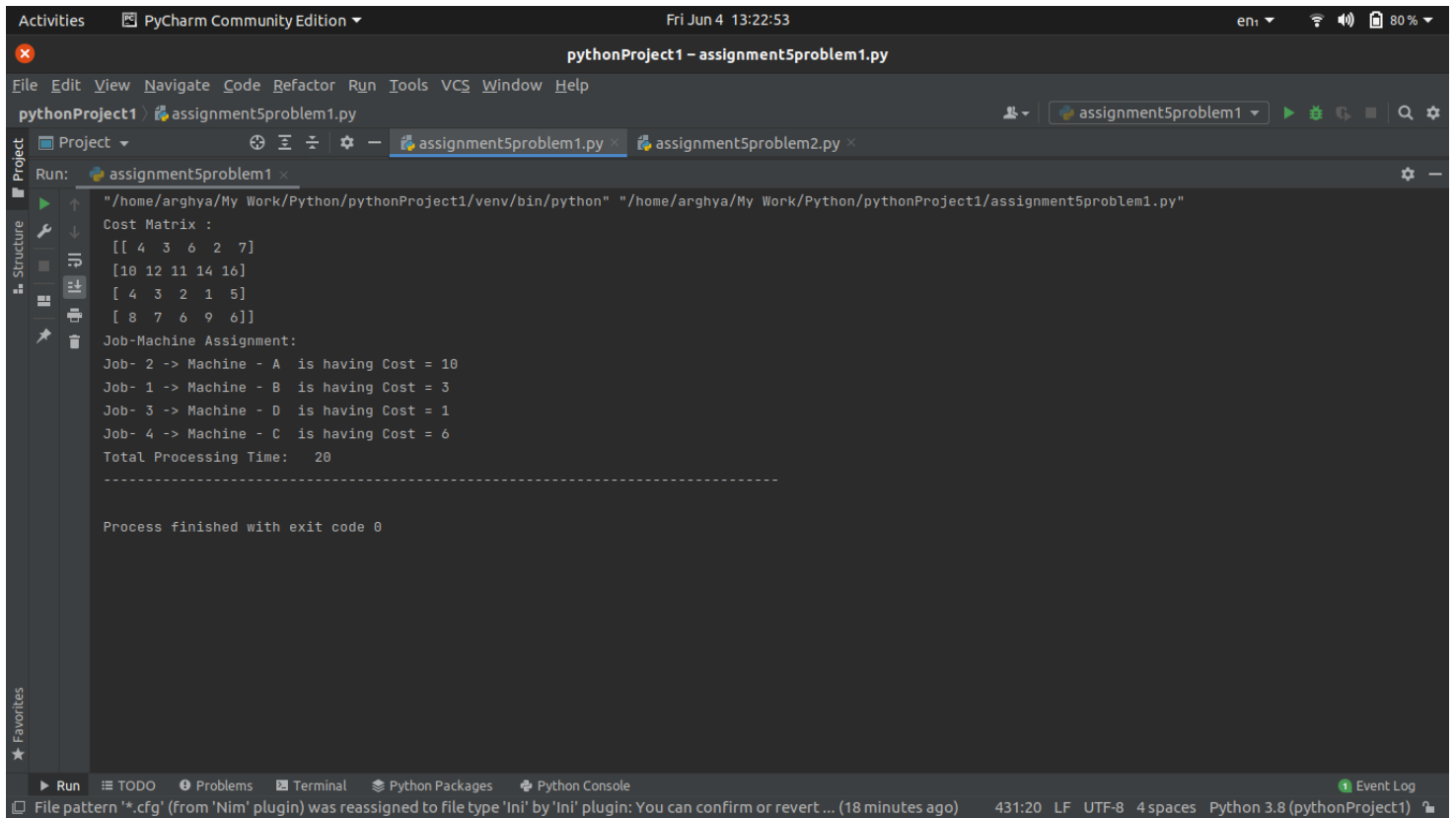
    cost_matrix = [
        [4, 3, 6, 2, 7],
        [10, 12, 11, 14, 16],
        [4, 3, 2, 1, 5],
        [8, 7, 6, 9, 6]]

    hungarian = Hungarian(cost_matrix)
    hungarian.calculate()
    print("Cost Matrix : \n", np.array(cost_matrix))
    jobs = [1, 2, 3, 4]
    machine = ['A', 'B', 'C', 'D', 'E']
    result = hungarian.get_results()
    print("Job-Machine Assignment:")
    for i in result:
        print("Job-", jobs[i[0]], "-> Machine -", machine[i[1]], " is having Cost =", c

```

```
print("Total Processing Time:\t", hungarian.get_total_potential())  
print("-" * 80)
```

Output :



The screenshot shows the PyCharm IDE interface. The top bar indicates the project is 'pythonProject1 - assignment5problem1.py' and the date is 'Fri Jun 4 13:22:53'. The main window displays the output of a Python script. The script's output is as follows:

```
"/home/arghya/My Work/Python/pythonProject1/venv/bin/python" "/home/arghya/My Work/Python/pythonProject1/assignment5problem1.py"
Cost Matrix :
[[ 4  3  6  2  7]
 [10 12 11 14 16]
 [ 4  3  2  1  5]
 [ 8  7  6  9  6]]
Job-Machine Assignment:
Job- 2 -> Machine - A  is having Cost = 10
Job- 1 -> Machine - B  is having Cost = 3
Job- 3 -> Machine - D  is having Cost = 1
Job- 4 -> Machine - C  is having Cost = 6
Total Processing Time:  20
-----

Process finished with exit code 0
```

The bottom status bar shows the file encoding as 'UTF-8' and the Python version as 'Python 3.8 (pythonProject1)'.

Problem statement 1

Find the optimal solution to the following transportation problem in which the cells contains the unit transportation cost in rupees.

	w_1	w_2	w_3	w_4	w_5	avail.
F_1	7	6	4	5	9	40
F_2	8	5	6	7	8	30
F_3	6	8	9	6	5	20
F_4	5	7	7	8	6	10
Req.	30	30	15	20	5	

use NWCR and LCM for initial basic feasible solution.

Problem formulation -

The capacity of F_2 and leaves $15 - 10 = 5$ units with w_3 .

The min value for $F_3 = 20$ and $w_3 = 5$ are compared. The $\min(20, 5) = 5$ is assigned to $F_3 w_3$. This meets the complete demand of w_3 and leaves $20 - 5 = 15$ units with F_3 .

The min values for $F_3 = 15$ and $w_4 = 20$ are compared.

The smaller of $15, 20 = 15$ is assigned for $F_3 w_4$, this exhaust the capacity of F_3 and leaves $20 - 15 = 5$ units with w_4 .

The min value for $F_4 = 10$ of $w_4 = 5$ are compared. The

$\min(10, 5) = 5$ is assign to $F_4 w_4$
 This meets the complete demand
 of w_4 and leaves $10 - 5 = 5$ units
 with F_4

The sum value for $F_4 = 5$ & $w_4 = 5$
 are compared.

The smaller of the two i.e.
 $\min(5, 5) = 5$ is assign to
 $F_4 w_5$

The TBFs =

	w_1	w_2	w_3	w_4	w_5	Supp.
F_1	7 (30)	6 (10)	4	5	9	40
F_2	8	5 (20)	6 (10)	7	8	30
F_3	6	8	9 (5)	6 (15)	5	20
F_4	5	7	7	8 (5)	6 (5)	10
Demand	30	30	15	20	5	

$$V_1 = 7 - 0 = 7$$

$$V_2 = 6 - 0 = 6$$

$$V_1 = C_{11} - u_1 = 7 - 0 \Rightarrow V_1 = 7$$

$$V_2 = C_{12} - u_1 = 6 - 0 \Rightarrow V_2 = 6$$

$$u_2 = C_{22} - V_2 = 5 - 6 \Rightarrow u_2 = -1$$

$$V_3 = C_{23} - u_2 = 6 + 1 \Rightarrow V_3 = 7$$

$$u_3 = C_{33} - V_3 = 9 - 7 \Rightarrow u_3 = 2$$

$$u_4 = C_{44} - V_4 = 8 - 4 \Rightarrow u_4 = 4$$

$$V_5 = C_{45} - u_4 = 6 - 4 \Rightarrow V_5 = 2$$

	w_1	w_2	w_3	w_4	w_5	S	u_i
F_1	7 ⁽²⁰⁾	6 ⁽¹⁰⁾	4	5	9	40	$u_1 = 0$
F_2	8	5 ⁽²⁰⁾	6 ⁽¹⁰⁾	7	8	20	$u_2 = -1$
F_3	6	8	9 ⁽⁵⁾	6 ⁽¹⁵⁾	5	20	$u_3 = 2$
F_4	5	7	7	8 ⁽⁵⁾	6 ⁽⁵⁾	10	$u_4 = 4$
F_5	30	30	15	20	5		
V_i	$V_1 = 7$	$V_2 = 6$	$V_3 = 7$	$V_4 = 4$	$V_5 = 2$		

The minimization T.P cost =

$$7 \times 30 + 6 \times 10 + 5 \times 20 + 8 \times 10 + 9 \times 5 \\ + 6 \times 15 + 8 \times 5 + 6 \times 5 = 635$$

$$\text{Total allocated cell} = 8 = m+n-1 \\ = 4+5-1 \\ = 8$$

hence the soln is non degenerate

optimality test using mauli method.

allocation table is

	w_1	w_2	w_3	w_4	w_5	Supp.
F_1	7 ⁽²⁰⁾	6 ⁽¹⁰⁾	9	5	9	40
F_2	8	5 ⁽²⁰⁾	6 ⁽¹⁰⁾	7	8	30
F_3	6	8	9 ⁽⁵⁾	6 ⁽¹⁵⁾	5	20
F_4	5	7	7	8 ⁽⁵⁾	6 ⁽⁵⁾	10
Dem.	30	30	15	20	5	

Iteration 1 of optimality test.

① since $u_i = 0$, we get

② find $d_{ij} = c_{ij} - (u_i + v_j)$

$$d_{13} = c_{13} - (u_1 + v_3) = 4 - (0 + 7) = -3$$

$$d_{14} = 5 - (0 + 4) = 1$$

$$d_{15} = 9 - (0 + 2) = 7$$

$$d_{21} = 8 - (-1 + 7) = 2$$

$$d_{24} = 7 - (-1 + 4) = 4$$

$$d_{25} = 8 - (-1 + 2) = 7$$

$$d_{31} = 6 - (2 + 7) = -3$$

$$d_{32} = 8 - (2 + 6) = 0$$

$$d_{35} = 5 - (2 + 2) = 1$$

$$d_{41} = 5 - (4 + 7) = -6$$

$$d_{42} = 7 - (4 + 6) = -3$$

$$d_{43} = 7 - (4 + 7) = -4$$

3)

	w_1	w_2	w_3	w_4	w_5	supp	u_i
f_1	7 ⁽⁻⁾	8 ⁽⁺⁾	4 ⁽⁻⁾	5	9 ⁽⁺⁾	40	0
f_2	8 ⁽⁺⁾	5 ⁽⁻⁾	6 ⁽⁺⁾	7 ⁽⁺⁾	8 ⁽⁺⁾	30	7
f_3	6 ⁽⁻⁾	8 ⁽⁺⁾	9 ⁽⁻⁾	6 ⁽⁺⁾	5 ⁽⁺⁾	20	2
f_4	5 ⁽⁺⁾	7 ⁽⁻⁾	7 ⁽⁺⁾	8 ⁽⁻⁾	6 ⁽⁺⁾	10	4
dem	30	30	15	20	5		
v_0	7	8	7	4	2		

Python Code:

```
class HungarianError(Exception):
    pass

try:
    import numpy as np
except ImportError:
    raise HungarianError("NumPy is not installed.")

class Hungarian:
    """
    Implementation of the Hungarian (Munkres) Algorithm using np.
    Usage:
        hungarian = Hungarian(cost_matrix)
        hungarian.calculate()
    or
        hungarian = Hungarian()
        hungarian.calculate(cost_matrix)
    Handle Profit matrix:
        hungarian = Hungarian(profit_matrix, is_profit_matrix=True)
    or
        cost_matrix = Hungarian.make_cost_matrix(profit_matrix)
    The matrix will be automatically padded if it is not square.
    For that numpy's resize function is used, which automatically adds 0's to any row/c
    Get results and total potential after calculation:
        hungarian.get_results()
        hungarian.get_total_potential()
    """

    def __init__(self, input_matrix=None, is_profit_matrix=False):
        """
        input_matrix is a List of Lists.
        input_matrix is assumed to be a cost matrix unless is_profit_matrix is True.
        """
        if input_matrix is not None:
            # Save input
            my_matrix = np.array(input_matrix)
            self._input_matrix = np.array(input_matrix)
            self._maxColumn = my_matrix.shape[1]
            self._maxRow = my_matrix.shape[0]

            # Adds 0s if any columns/rows are added. Otherwise stays unaltered
            matrix_size = max(self._maxColumn, self._maxRow)
            pad_columns = matrix_size - self._maxRow
            pad_rows = matrix_size - self._maxColumn
            my_matrix = np.pad(my_matrix, ((0, pad_columns), (0, pad_rows)), 'constant')

            # Convert matrix to profit matrix if necessary
            if is_profit_matrix:
                my_matrix = self.make_cost_matrix(my_matrix)
```

```

        self._cost_matrix = my_matrix
        self._size = len(my_matrix)
        self._shape = my_matrix.shape

        # Results from algorithm.
        self._results = []
        self._totalPotential = 0
    else:
        self._cost_matrix = None

def get_results(self):
    """Get results after calculation."""
    return self._results

def get_total_potential(self):
    """Returns expected value after calculation."""
    return self._totalPotential

def calculate(self, input_matrix=None, is_profit_matrix=False):
    """
    Implementation of the Hungarian (Munkres) Algorithm.
    input_matrix is a List of Lists.
    input_matrix is assumed to be a cost matrix unless is_profit_matrix is True.
    """
    # Handle invalid and new matrix inputs.
    if input_matrix is None and self._cost_matrix is None:
        raise HungarianError("Invalid input")
    elif input_matrix is not None:
        self._init__(input_matrix, is_profit_matrix)

    result_matrix = self._cost_matrix.copy()

    # Step 1: Subtract row mins from each row.
    for index, row in enumerate(result_matrix):
        result_matrix[index] -= row.min()

    # Step 2: Subtract column mins from each column.
    for index, column in enumerate(result_matrix.T):
        result_matrix[:, index] -= column.min()

    # Step 3: Use minimum number of lines to cover all zeros in the matrix.
    # If the total covered rows+columns is not equal to the matrix size then adjust
    total_covered = 0
    while total_covered < self._size:
        # Find minimum number of lines to cover all zeros in the matrix and find to
        cover_zeros = CoverZeros(result_matrix)
        covered_rows = cover_zeros.get_covered_rows()
        covered_columns = cover_zeros.get_covered_columns()
        total_covered = len(covered_rows) + len(covered_columns)

    # if the total covered rows+columns is not equal to the matrix size then adjust
    if total_covered < self._size:
        result_matrix = self._adjust_matrix_by_min_uncovered_num(result_matrix,

```

```

# Step 4: Starting with the top row, work your way downwards as you make assignments
# Find single zeros in rows or columns.
# Add them to final result and remove them and their associated row/column from
expected_results = min(self._maxColumn, self._maxRow)
zero_locations = (result_matrix == 0)
while len(self._results) != expected_results:

    # If number of zeros in the matrix is zero before finding all the results then
    if not zero_locations.any():
        raise HungarianError("Unable to find results. Algorithm has failed.")

    # Find results and mark rows and columns for deletion
    matched_rows, matched_columns = self._find_matches(zero_locations)

    # Make arbitrary selection
    total_matched = len(matched_rows) + len(matched_columns)
    if total_matched == 0:
        matched_rows, matched_columns = self.select_arbitrary_match(zero_locations)

    # Delete rows and columns
    for row in matched_rows:
        zero_locations[row] = False
    for column in matched_columns:
        zero_locations[:, column] = False

    # Save Results
    self._set_results(zip(matched_rows, matched_columns))

# Calculate total potential
value = 0
for row, column in self._results:
    value += self._input_matrix[row, column]
self._totalPotential = value

@staticmethod
def make_cost_matrix(profit_matrix):
    """
    Converts a profit matrix into a cost matrix.
    Expects NumPy objects as input.
    """
    # subtract profit matrix from a matrix made of the max value of the profit matrix
    matrix_shape = profit_matrix.shape
    offset_matrix = np.ones(matrix_shape, dtype=int) * profit_matrix.max()
    cost_matrix = offset_matrix - profit_matrix
    return cost_matrix

def _adjust_matrix_by_min_uncovered_num(self, result_matrix, covered_rows, covered_columns):
    """ Subtract m from every uncovered number and add m to every element covered with m
    # Calculate minimum uncovered number (m)
    elements = []
    for row_index, row in enumerate(result_matrix):
        if row_index not in covered_rows:
            for index, element in enumerate(row):
                if index not in covered_columns:

```



```

        elements.append(element)
min_uncovered_num = min(elements)

# Add m to every covered element
adjusted_matrix = result_matrix
for row in covered_rows:
    adjusted_matrix[row] += min_uncovered_num
for column in covered_columns:
    adjusted_matrix[:, column] += min_uncovered_num

# Subtract m from every element
m_matrix = np.ones(self._shape, dtype=int) * min_uncovered_num
adjusted_matrix -= m_matrix

return adjusted_matrix

def __find_matches(self, zero_locations):
    """Returns rows and columns with matches in them."""
    marked_rows = np.array([], dtype=int)
    marked_columns = np.array([], dtype=int)

    # Mark rows and columns with matches
    # Iterate over rows
    for index, row in enumerate(zero_locations):
        row_index = np.array([index])
        if np.sum(row) == 1:
            column_index, = np.where(row)
            marked_rows, marked_columns = self.__mark_rows_and_columns(marked_rows,
                                                                       column_index)

    # Iterate over columns
    for index, column in enumerate(zero_locations.T):
        column_index = np.array([index])
        if np.sum(column) == 1:
            row_index, = np.where(column)
            marked_rows, marked_columns = self.__mark_rows_and_columns(marked_rows,
                                                                       column_index)

    return marked_rows, marked_columns

    @staticmethod
def __mark_rows_and_columns(marked_rows, marked_columns, row_index, column_index):
    """Check if column or row is marked. If not marked then mark it."""
    new_marked_rows = marked_rows
    new_marked_columns = marked_columns
    if not (marked_rows == row_index).any() and not (marked_columns == column_index):
        new_marked_rows = np.insert(marked_rows, len(marked_rows), row_index)
        new_marked_columns = np.insert(marked_columns, len(marked_columns), column_index)
    return new_marked_rows, new_marked_columns

    @staticmethod
def select_arbitrary_match(zero_locations):
    """Selects row column combination with minimum number of zeros in it."""
    # Count number of zeros in row and column combinations

```

```

rows, columns = np.where(zero_locations)
zero_count = []
for index, row in enumerate(rows):
    total_zeros = np.sum(zero_locations[row]) + np.sum(zero_locations[:, columns[index]])
    zero_count.append(total_zeros)

# Get the row column combination with the minimum number of zeros.
indices = zero_count.index(min(zero_count))
row = np.array([rows[indices]])
column = np.array([columns[indices]])

return row, column

def __set_results(self, result_lists):
    """Set results during calculation."""
    # Check if results values are out of bound from input matrix (because of matrix
    # Add results to results list.
    for result in result_lists:
        row, column = result
        if row < self._maxRow and column < self._maxColumn:
            new_result = (int(row), int(column))
            self._results.append(new_result)

class CoverZeros:
    """
    Use minimum number of lines to cover all zeros in the matrix.
    Algorithm based on: http://weber.ucsd.edu/~vcrawfor/hungar.pdf
    """

    def __init__(self, matrix):
        """
        Input a matrix and save it as a boolean matrix to designate zero locations.
        Run calculation procedure to generate results.
        """
        # Find zeros in matrix
        self._zero_locations = (matrix == 0)
        self._shape = matrix.shape

        # Choices starts without any choices made.
        self._choices = np.zeros(self._shape, dtype=bool)

        self._marked_rows = []
        self._marked_columns = []

        # marks rows and columns
        self._calculate()

        # Draw lines through all unmarked rows and all marked columns.
        self._covered_rows = list(set(range(self._shape[0])) - set(self._marked_rows))
        self._covered_columns = self._marked_columns

    def get_covered_rows(self):
        """Return list of covered rows."""

```

```

        return self._covered_rows

def get_covered_columns(self):
    """Return list of covered columns."""
    return self._covered_columns

def __calculate(self):
    """
    Calculates minimum number of lines necessary to cover all zeros in a matrix.
    Algorithm based on: http://weber.ucsd.edu/~vcrawfor/hungar.pdf
    """
    while True:
        # Erase all marks.
        self._marked_rows = []
        self._marked_columns = []

        # Mark all rows in which no choice has been made.
        for index, row in enumerate(self._choices):
            if not row.any():
                self._marked_rows.append(index)

        # If no marked rows then finish.
        if not self._marked_rows:
            return True

        # Mark all columns not already marked which have zeros in marked rows.
        num_marked_columns = self.__mark_new_columns_with_zeros_in_marked_rows()

        # If no new marked columns then finish.
        if num_marked_columns == 0:
            return True

        # While there is some choice in every marked column.
        while self.__choice_in_all_marked_columns():
            # Some Choice in every marked column.

            # Mark all rows not already marked which have choices in marked columns
            num_marked_rows = self.__mark_new_rows_with_choices_in_marked_columns()

            # If no new marks then Finish.
            if num_marked_rows == 0:
                return True

            # Mark all columns not already marked which have zeros in marked rows.
            num_marked_columns = self.__mark_new_columns_with_zeros_in_marked_rows()

            # If no new marked columns then finish.
            if num_marked_columns == 0:
                return True

        # No choice in one or more marked columns.
        # Find a marked column that does not have a choice.
        choice_column_index = self.__find_marked_column_without_choice()

```

```

while choice_column_index is not None:
    # Find a zero in the column indexed that does not have a row with a cho
    choice_row_index = self._find_row_without_choice(choice_column_index)

    # Check if an available row was found.
    new_choice_column_index = None
    if choice_row_index is None:
        # Find a good row to accomodate swap. Find its column pair.
        choice_row_index, new_choice_column_index = self._find_best_choice_r
        choice_column_index)

    # Delete old choice.
    self._choices[choice_row_index, new_choice_column_index] = False

    # Set zero to choice.
    self._choices[choice_row_index, choice_column_index] = True

    # Loop again if choice is added to a row with a choice already in it.
    choice_column_index = new_choice_column_index

def _mark_new_columns_with_zeros_in_marked_rows(self):
    """Mark all columns not already marked which have zeros in marked rows."""
    num_marked_columns = 0
    for index, column in enumerate(self._zero_locations.T):
        if index not in self._marked_columns:
            if column.any():
                row_indices, = np.where(column)
                zeros_in_marked_rows = (set(self._marked_rows) & set(row_indices))
                if zeros_in_marked_rows:
                    self._marked_columns.append(index)
                    num_marked_columns += 1
    return num_marked_columns

def _mark_new_rows_with_choices_in_marked_columns(self):
    """Mark all rows not already marked which have choices in marked columns."""
    num_marked_rows = 0
    for index, row in enumerate(self._choices):
        if index not in self._marked_rows:
            if row.any():
                column_index, = np.where(row)
                if column_index in self._marked_columns:
                    self._marked_rows.append(index)
                    num_marked_rows += 1
    return num_marked_rows

def _choice_in_all_marked_columns(self):
    """Return Boolean True if there is a choice in all marked columns. Returns bool
    for column_index in self._marked_columns:
        if not self._choices[:, column_index].any():
            return False
    return True

def _find_marked_column_without_choice(self):
    """Find a marked column that does not have a choice."""

```

```

        for column_index in self._marked_columns:
            if not self._choices[:, column_index].any():
                return column_index

        raise HungarianError(
            "Could not find a column without a choice. Failed to cover matrix zeros. All

def __find_row_without_choice(self, choice_column_index):
    """Find a row without a choice in it for the column indexed. If a row does not
    row_indices, = np.where(self._zero_locations[:, choice_column_index])
    for row_index in row_indices:
        if not self._choices[row_index].any():
            return row_index

    # All rows have choices. Return None.
    return None

def __find_best_choice_row_and_new_column(self, choice_column_index):
    """
    Find a row index to use for the choice so that the column that needs to be changed.
    Return a random row and column if unable to find an optimal selection.
    """
    row_indices, = np.where(self._zero_locations[:, choice_column_index])
    for row_index in row_indices:
        column_indices, = np.where(self._choices[row_index])
        column_index = column_indices[0]
        if self.__find_row_without_choice(column_index) is not None:
            return row_index, column_index

    # Cannot find optimal row and column. Return a random row and column.
    from random import shuffle

    shuffle(row_indices)
    column_index, = np.where(self._choices[row_indices[0]])
    return row_indices[0], column_index[0]

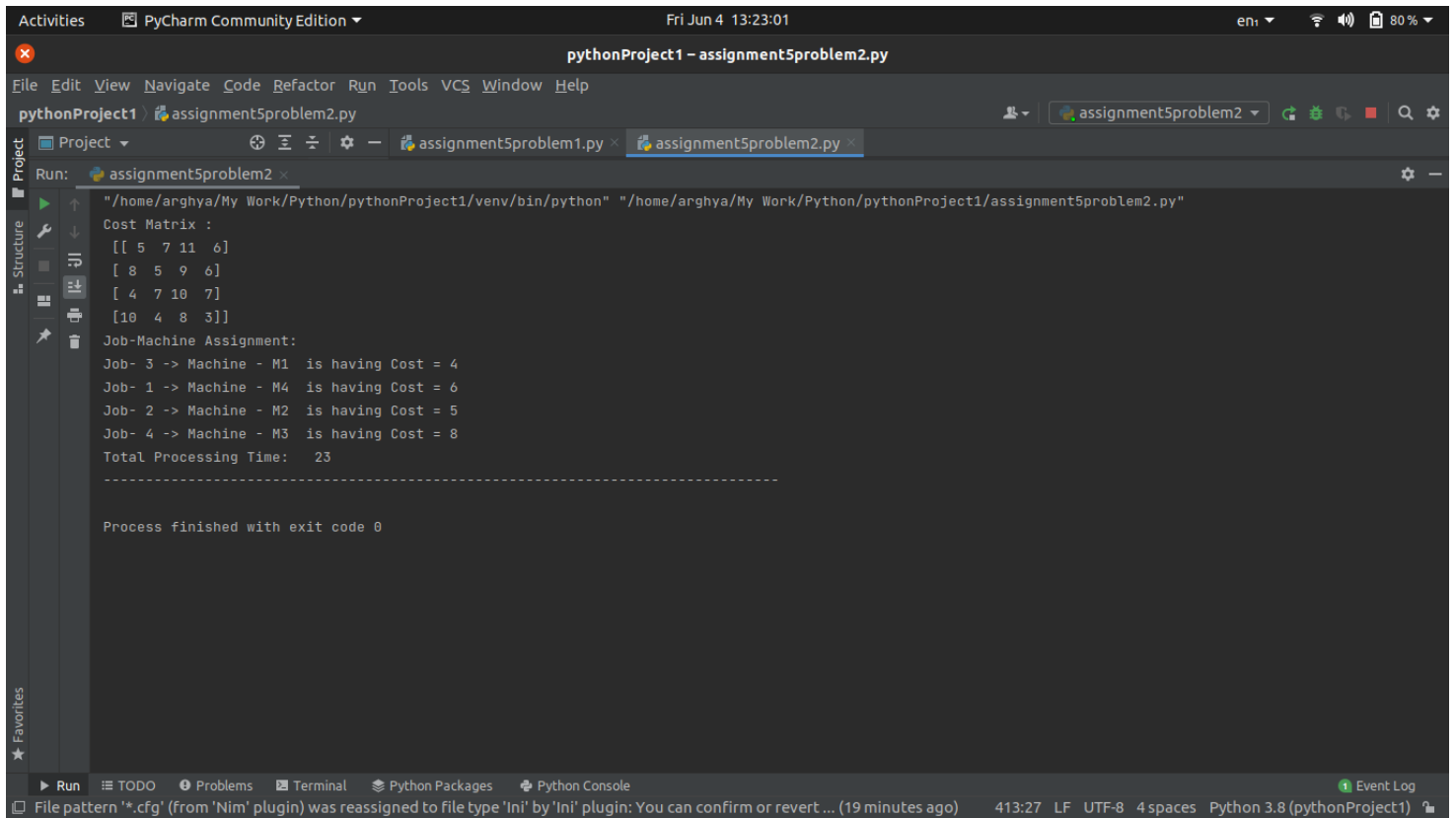
if __name__ == '__main__':
    cost_matrix = [
        [5, 7, 11, 6],
        [8, 5, 9, 6],
        [4, 7, 10, 7],
        [10, 4, 8, 3]]

    hungarian = Hungarian(cost_matrix)
    hungarian.calculate()
    print("Cost Matrix : \n", np.array(cost_matrix))
    jobs = [1, 2, 3, 4]
    machine = ['M1', 'M2', 'M3', 'M4']
    result = hungarian.get_results()
    print("Job-Machine Assignment:")
    for i in result:
        print("Job-", jobs[i[0]], "-> Machine -", machine[i[1]], " is having Cost =", cost_matrix[i[0]][i[1]])
    print("Total Processing Time:\t", hungarian.get_total_potential())

```

```
print("-" * 80)
```

Output :



The screenshot shows the PyCharm IDE interface. The top bar indicates the project is 'pythonProject1 - assignment5problem2.py'. The left sidebar shows the 'Run' tab selected. The main window displays the output of the script 'assignment5problem2.py'.

```
Run: assignment5problem2
"/home/arghya/My Work/Python/pythonProject1/venv/bin/python" "/home/arghya/My Work/Python/pythonProject1/assignment5problem2.py"
Cost Matrix :
[[ 5  7 11  6]
 [ 8  5  9  6]
 [ 4  7 10  7]
 [10  4  8  3]]
Job-Machine Assignment:
Job- 3 -> Machine - M1 is having Cost = 4
Job- 1 -> Machine - M4 is having Cost = 6
Job- 2 -> Machine - M2 is having Cost = 5
Job- 4 -> Machine - M3 is having Cost = 8
Total Processing Time: 23
-----

Process finished with exit code 0
```

The bottom status bar shows the file encoding as UTF-8, 4 spaces, and Python 3.8 (pythonProject1).