# LP 3 Practicals DAA

Practical 1:- Write a program non-recursive and recursive program to calculate Fibonacci numbers and analyze their time and space complexity.

```
def recur(n):
  if n <= 1:
    return n
  else:
    return(recur(n-1) + recur(n-2))
def iterative(n):
  a = 0
  b = 1
  print(a)
  print(b)
  for i in range(2, n):
    print(a + b)
    a, b = b, a + b
if __name___== "__main___":
  num = int(input("Enter the nth number for series: "))
  if num <= 0:
    print("Please enter a positive integer")
  else:
    print("Fibonacci sequence with recursion:")
    for i in range(num):
       print(recur(i))
  print("Fibonacci series with Iteration:")
  iterative(num)
```

#### Practical 2:- Write a program to solve a fractional Knapsack problem using a greedy method.

```
class Item:
  def init (self, value, weight):
    self.value = value
    self.weight = weight
def fractional_knapsack(capacity, items):
  items.sort(key=lambda item: item.value/item.weight, reverse=True)
  total_value = 0
  for item in items:
    if capacity >= item.weight:
      capacity -= item.weight
      total_value += item.value
    else:
      fraction = capacity / item.weight
      total value += item.value * fraction
      break
  return total_value
if name == " main ":
  n = int(input("Enter the number of items: "))
  items = []
  for i in range(n):
    value = float(input(f"Enter value of item {i+1}: "))
    weight = float(input(f"Enter weight of item {i+1}: "))
    items.append(Item(value, weight))
  capacity = float(input("Enter the capacity of the knapsack: "))
  max_value = fractional_knapsack(capacity, items)
  print(f"Maximum value we can obtain = {max_value}")
```

### Practical 3:- Write a program to solve a 0-1 Knapsack problem using dynamic programming or branch and bound strategy.

#### 1. Knapsack problem using dynamic

total weight += items[j].weight

```
def knapsack_01(weights, values, capacity):
  n = len(values)
  dp = [[0 \text{ for } x \text{ in range}(capacity + 1)] \text{ for } y \text{ in range}(n + 1)]
  for i in range(n + 1):
    for w in range(capacity + 1):
       if i == 0 or w == 0:
         dp[i][w] = 0
       elif weights[i - 1] <= w:
         dp[i][w] = max(values[i-1] + dp[i-1][w-weights[i-1]], dp[i-1][w])
         dp[i][w] = dp[i - 1][w]
  return dp[n][capacity]
if name == " main ":
  n = int(input("Enter the number of items: "))
  values = []
  weights = []
  for i in range(n):
    value = int(input(f"Enter value of item {i+1}: "))
    weight = int(input(f"Enter weight of item {i+1}: "))
    values.append(value)
    weights.append(weight)
  capacity = int(input("Enter the capacity of the knapsack: "))
  max_value = knapsack_01(weights, values, capacity)
  print(f"Maximum value we can obtain = {max_value}")
2. Branch and bound strategy
class Item:
  def __init_(self, value, weight):
    self.value = value
    self.weight = weight
    self.ratio = value / weight
class Node:
  def init (self, level, value, weight, bound):
    self.level = level
    self.value = value
    self.weight = weight
    self.bound = bound
def bound(node, capacity, n, items):
  if node.weight >= capacity:
    return 0
  profit bound = node.value
  j = node.level + 1
  total_weight = node.weight
  while j < n and total_weight + items[j].weight <= capacity:
```

```
profit bound += items[j].value
    j += 1
  if j < n:
    profit bound += (capacity - total weight) * items[j].ratio
  return profit bound
def knapsack_branch_and_bound(capacity, items):
  items.sort(key=lambda x: x.ratio, reverse=True)
  n = len(items)
  Q = []
  u = Node(-1, 0, 0, 0)
  v = Node(-1, 0, 0, 0)
  u.bound = bound(u, capacity, n, items)
  Q.append(u)
  \max profit = 0
  while Q:
    u = Q.pop(0)
    if u.level == -1:
      v.level = 0
    if u.level == n - 1:
      continue
    v.level = u.level + 1
    v.weight = u.weight + items[v.level].weight
    v.value = u.value + items[v.level].value
    if v.weight <= capacity and v.value > max profit:
      max_profit = v.value
    v.bound = bound(v, capacity, n, items)
    if v.bound > max_profit:
      Q.append(v)
    v.weight = u.weight
    v.value = u.value
    v.bound = bound(v, capacity, n, items)
    if v.bound > max_profit:
      Q.append(v)
  return max_profit
if __name___== "__main___":
  n = int(input("Enter the number of items: "))
  items = []
  for i in range(n):
    value = int(input(f"Enter value of item {i+1}: "))
    weight = int(input(f"Enter weight of item {i+1}: "))
    items.append(Item(value, weight))
  capacity = int(input("Enter the capacity of the knapsack: "))
  max_value = knapsack_branch_and_bound(capacity, items)
  print(f"Maximum value we can obtain = {max_value}")
```

# Practical 4:- Design n-Queens matrix having first Queen placed. Use backtracking to place remaining Queens to generate the final n-queen's matrix

```
def is safe(board, row, col, n):
  for i in range(col):
    if board[row][i] == 1:
       return False
  for i, j in zip(range(row, -1, -1), range(col, -1, -1)):
    if board[i][j] == 1:
       return False
  for i, j in zip(range(row, n, 1), range(col, -1, -1)):
    if board[i][j] == 1:
       return False
  return True
def solve_n_queens(board, col, n):
  if col >= n:
    return True
  for i in range(n):
    if is_safe(board, i, col, n):
       board[i][col] = 1
       if solve_n_queens(board, col + 1, n):
         return True
       board[i][col] = 0
  return False
def n_queens(n, start_row, start_col):
  board = [[0 for _ in range(n)] for _ in range(n)]
  board[start_row][start_col] = 1
  if solve n queens(board, 1, n): # Start from the next column
    return board
  else:
    return None
def print board(board):
  if board:
    for row in board:
       print("".join("Q" if x == 1 else "." for x in row))
    print("No solution found")
n = int(input("Enter the value of n (size of board): "))
start_row = 0
start col = 0
final_board = n_queens(n, start_row, start_col)
print_board(final_board)
```

#### Practical 5:- Write a program for analysis of quick sort by using deterministic and randomized variant

```
import time
def quick sort deterministic(arr, low, high):
  if low < high:
    pi = partition(arr, low, high)
    quick_sort_deterministic(arr, low, pi - 1)
    quick_sort_deterministic(arr, pi + 1, high)
def quick_sort_randomized(arr, low, high):
  if low < high:
    random index = low + (hash(str(arr)) % (high - low + 1)) # Simple random index generation
    arr[low], arr[random_index] = arr[random_index], arr[low]
    pi = partition(arr, low, high)
    quick sort randomized(arr, low, pi - 1)
    quick sort randomized(arr, pi + 1, high)
def partition(arr, low, high):
  pivot = arr[high]
  i = low - 1
  for j in range(low, high):
    if arr[j] <= pivot:
      i += 1
      arr[i], arr[j] = arr[j], arr[i]
  arr[i + 1], arr[high] = arr[high], arr[i + 1]
  return i + 1
def analyze_sorting_algorithm(sort_function, arr):
  start time = time.time()
  sort function(arr, 0, len(arr) - 1)
  end_time = time.time()
  return end_time - start_time
def generate random array(size):
  return [i for i in range(size, 0, -1)] # Simple reverse sorted array for worst-case
if __name___== "__main___":
  array size = int(input("Enter the size of the array: "))
  arr = generate random array(array size)
  arr deterministic = arr.copy()
  arr_randomized = arr.copy()
  print("Analyzing deterministic Quick Sort...")
  deterministic time = analyze sorting algorithm(quick sort deterministic, arr deterministic)
  print("Analyzing randomized Quick Sort...")
  randomized_time = analyze_sorting_algorithm(quick_sort_randomized, arr_randomized)
  print(f"Deterministic Quick Sort Time: {deterministic_time:.6f} seconds")
  print(f"Randomized Quick Sort Time: {randomized time:.6f} seconds")
```

Practical 1:- Predict the price of the Uber ride from a given pickup point to the agreed drop-off location. Perform following tasks:

- 1. Preprocess the dataset.
- 2. Identify outliers.
- 3. Check the correlation.
- 4. Implement linear regression and random forest regression models.
- 5. Evaluate the models and compare their respective scores like R2, RMSE, etc.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
from sklearn.model_selection import train_test_split
from sklearn.linear model import LinearRegression
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error, r2_score
url = "uber.csv"
df = pd.read csv(url)
print(df.head())
1. Data Preprocessing
print(df.isnull().sum())
df.dropna(inplace=True)
df['pickup_datetime'] = pd.to_datetime(df['pickup_datetime'])
df['hour'] = df['pickup datetime'].dt.hour
df['day of week'] = df['pickup datetime'].dt.dayofweekdf = df[['passenger count',
'pickup_longitude', 'pickup_latitude', 'dropoff_longitude', 'dropoff_latitude', 'hour', 'fare_amount']]
2. Identify Outliers
z scores = np.abs(stats.zscore(df['fare amount']))
df.loc[:, 'outlier'] = z_scores > 3
outliers = df[df['outlier']]
print("Number of outliers detected:", len(outliers))
3. Check Correlation
correlation matrix = df.corr()
print(correlation_matrix)
plt.figure(figsize=(10, 8))
sns.heatmap(correlation matrix, annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Correlation Matrix')
plt.show()
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
4. Implement Random Forest Regression
linear model = LinearRegression()
```

```
linear_model.fit(X_train, y_train)
y_pred_linear = linear_model.predict(X_test)

5. Evaluate Models
rf_model = RandomForestRegressor(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)
y_pred_rf = rf_model.predict(X_test)

6. Evaluate Models
r2_linear = r2_score(y_test, y_pred_linear)
rmse_linear = np.sqrt(mean_squared_error(y_test, y_pred_linear))
r2_rf = r2_score(y_test, y_pred_rf)
rmse_rf = np.sqrt(mean_squared_error(y_test, y_pred_rf))
print(f'Linear Regression: R² = {r2_linear:.4f}, RMSE = {rmse_linear:.4f}')
print(f'Random Forest Regression: R² = {r2_rf:.4f}, RMSE = {rmse_rf:.4f}')
```

### Practical 2:- Classify the email using the binary classification method. Email Spam detection has two states:

```
a) Normal State - Not Spam
```

b) Abnormal State – Spam. Use K-Nearest Neighbors and Support Vector Machine for classification. Analyze their performance.

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
Step 1: Load the dataset
url = "emails.csv"
df = pd.read csv('emails.csv', encoding='latin-1')
print(df.head())
Step 2: Data Preprocessing
X = df.drop(columns=['Email No.', 'Prediction'])
y = df['Prediction']
X = X.apply(pd.to_numeric, errors='coerce')
X = X.dropna()
y = y[X.index]
Step 3: Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
Step 4: K-Nearest Neighbors Model
knn model = KNeighborsClassifier(n neighbors=5)
knn_model.fit(X_train, y_train)
y_pred_knn = knn_model.predict(X_test)
Step 5: Support Vector Machine Model
svm_model = SVC(kernel='linear')
svm_model.fit(X_train, y_train)
y_pred_svm = svm_model.predict(X_test)
Step 6: Evaluate Models
print("K-Nearest Neighbors Classification Report:")
print(classification_report(y_test, y_pred_knn, target_names=['Normal State - Not Spam (0)',
'Abnormal State - Spam (1)']))
print("Confusion Matrix:")
print(confusion matrix(y test, y pred knn))
print("Accuracy:", accuracy_score(y_test, y_pred_knn))
print("\nSupport Vector Machine Classification Report:")
print(classification_report(y_test, y_pred_svm, target_names=['Normal State - Not Spam (0)',
'Abnormal State - Spam (1)']))
print("Confusion Matrix:")
print(confusion_matrix(y_test, y_pred_svm))
print("Accuracy:", accuracy_score(y_test, y_pred_svm))
```

## Practical 3:- Given a bank customer, build a neural network-based classifier that can determine whether they will leave or not in the next 6 months

```
import pandas as pd
import numpy as np
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score, confusion_matrix
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.utils import to_categorical
# Step 1: Read the dataset
url = 'Churn Modelling.csv'
df = pd.read csv(url)
# Step 2: Distinguish feature and target set
X = df.drop(columns=['RowNumber', 'CustomerId', 'Surname', 'Exited'])
y = df['Exited']
X = pd.get_dummies(X, drop_first=True)
# Step 3: Split the dataset into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
scaler = StandardScaler()
X train = scaler.fit transform(X train)
X_test = scaler.transform(X_test)
# Step 4: Initialize and build the model
model = Sequential()
model.add(Dense(32, activation='relu', input shape=(X train.shape[1],))) # First hidden layer
model.add(Dense(16, activation='relu')) # Second hidden layer
model.add(Dense(1, activation='sigmoid')) # Output layer for binary classification
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
model.fit(X_train, y_train, epochs=50, batch_size=32, validation_split=0.2)
# Step 5: Evaluate the model
y_pred = model.predict(X_test)
y pred = (y pred > 0.5).astype(int)
accuracy = accuracy score(y test, y pred)
conf_matrix = confusion_matrix(y_test, y_pred)
print(f'Accuracy: {accuracy:.4f}')
print('Confusion Matrix:')
print(conf_matrix)
```

# Practical 4:- Implement Gradient Descent Algorithm to find the local minima of a function. For example, find the local minima of the function $y=(x+3)^2$ starting from the point x=2.

```
import numpy as np
import matplotlib.pyplot as plt
# Step 1: Define the function and its derivative
def function(x):
  return (x + 3)**2
def derivative(x):
  return 2 * (x + 3)
# Step 2: Implement the Gradient Descent Algorithm
def gradient_descent(starting_point, learning_rate, num_iterations):
  x = starting_point
  x_history = [x]
  for in range(num iterations):
    x -= learning_rate * derivative(x)
    x_history.append(x)
  return x, x_history
# Step 3: Parameters for Gradient Descent
starting_point = 2
learning_rate = 0.1
num_iterations = 50
local minima, history = gradient descent(starting point, learning rate, num iterations)
print(f'Local minima found at x = {local_minima:.4f}, y = {function(local_minima):.4f}')
# Step 4: Visualizing the results
x vals = np.linspace(-6, 0, 100)
y vals = function(x vals)
plt.plot(x_vals, y_vals, label='y = (x + 3)^{2}')
plt.scatter(history, function(np.array(history)), color='red', label='Gradient Descent Steps')
plt.xlabel('x')
plt.ylabel('y')
plt.title('Gradient Descent to Find Local Minima')
plt.axhline(0, color='black', linewidth=0.5, ls='--')
plt.axvline(0, color='black', linewidth=0.5, ls='--')
plt.legend()
plt.grid()
plt.show()
```

### Practical 5:- Implement K-Means clustering/ hierarchical clustering on sales\_data\_sample.csv dataset. Determine the number of clusters using the elbow method.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
from scipy.cluster.hierarchy import dendrogram, linkage
import seaborn as sns
# Step 1: Load the dataset
url = "sales data sample.csv"
df = pd.read csv(url, encoding='ISO-8859-1')
# Step 2: Preprocess the data
features = df[['QUANTITYORDERED', 'PRICEEACH']].copy()
features.dropna(inplace=True)
scaler = StandardScaler()
features_scaled = scaler.fit_transform(features)
# Step 3: Determine the number of clusters using the elbow method
wcss = []
for i in range(1, 11):
  kmeans = KMeans(n clusters=i, random state=42)
  kmeans.fit(features scaled)
  wcss.append(kmeans.inertia)
plt.figure(figsize=(10, 6))
plt.plot(range(1, 11), wcss, marker='o')
plt.title('Elbow Method for Optimal k')
plt.xlabel('Number of clusters (k)')
plt.ylabel('WCSS')
plt.xticks(range(1, 11))
plt.grid()
plt.show()
# Step 4: Apply K-Means clustering
optimal_k = 3
kmeans = KMeans(n clusters=optimal k, random state=42)
df['Cluster'] = kmeans.fit predict(features scaled)
plt.figure(figsize=(10, 6))
plt.scatter(df['QUANTITYORDERED'], df['PRICEEACH'], c=df['Cluster'], cmap='viridis')
plt.title('K-Means Clustering Results')
plt.xlabel('QUANTITYORDERED')
plt.ylabel('PRICEEACH')
plt.grid()
plt.colorbar(label='Cluster')
plt.show()
# Step 5: (Optional) Hierarchical Clustering
linkage_matrix = linkage(features_scaled, method='ward')
plt.figure(figsize=(12, 8))
```

```
dendrogram(linkage_matrix)
plt.title('Dendrogram for Hierarchical Clustering')
plt.xlabel('Samples')
plt.ylabel('Distance')
plt.grid()
plt.show()
```

Practical 1:- Write a smart contract on a test network, for Bank account of a customer for following operations:

```
• Deposit money

    Withdraw Money

    Show balance

// SPDX-License-Identifier: MIT
pragma solidity ^0.8.0;
contract BankAccount {
  mapping(address => uint256) private balances;
  event Deposit(address indexed account, uint256 amountInEther);
  event Withdraw(address indexed account, uint256 amountInEther);
  function deposit() public payable {
    uint256 amountInEther = msg.value / 1 ether;
    require(amountInEther > 0, "Deposit amount should be greater than 0");
    balances[msg.sender] += amountInEther;
    emit Deposit(msg.sender, amountInEther);
  function withdraw(uint256 amountInEther) public {
    require(amountInEther <= balances[msg.sender], "Insufficient balance");
    balances[msg.sender] -= amountInEther;
    uint256 amountInWei = amountInEther * 1 ether;
    payable(msg.sender).transfer(amountInWei);
    emit Withdraw(msg.sender, amountInEther);
  function showBalance() public view returns (uint256) {
    return balances[msg.sender];
```

}

#### Practical 2:- Write a program in solidity to create Student data. Use the following constructs:

- Structures
- Arrays
- Fallback

```
Deploy this as smart contract on Ethereum and Observe the transaction fee and Gas values.
```

```
// SPDX-License-Identifier: MIT
//https://betterprogramming.pub/developing-a-smart-contract-by-using-re mix-ide-81ff6f44ba2f
pragma solidity ^0.5.0;
contract Crud {
  struct User {
    uint id;
    string name;
  User[] public users;
  uint public nextId = 0;
  function Create(string memory name) public {
    users.push(User(nextId, name));
    nextId++;
  function Read(uint id) view public returns(uint, string memory) {
    for(uint i=0; i<users.length; i++) {</pre>
       if(users[i].id == id) {
         return(users[i].id, users[i].name);
       }
    }
  }
  function Update(uint id, string memory name) public {
    for(uint i=0; i<users.length; i++) {
       if(users[i].id == id) {
         users[i].name =name;
       }
    }
  }
  function Delete(uint id) public {
    delete users[id];
  function find(uint id) view internal returns(uint) {
    for(uint i=0; i< users.length; i++) {
       if(users[i].id == id) {
         return i;
       }
    // if user does not exist then revert back
    revert("User does not exist");
  }
}
```

#### Practical No :3

```
// SPDX-License-Identifier: MIT
pragma solidity ^0.8.0;
contract Crud {
    struct User {
        uint id;
        string name;
    }
    User[] public users;
    uint public nextId = 0;
    function Create(string memory name) public {
        users.push(User(nextId, name));
        nextId++;
    }
    function Read(uint id) view public returns(uint, string memory) {
        for(uint i=0; i<users.length; i++) {</pre>
            if(users[i].id == id) {
                return(users[i].id, users[i].name);
        }
    function Update(uint id, string memory name) public {
        for(uint i=0; i<users.length; i++) {</pre>
            if(users[i].id == id) {
                users[i].name =name;
        }
    function Delete(uint id) public {
        delete users[id];
    function find(uint id) view internal returns(uint) {
        for(uint i=0; i< users.length; i++) {</pre>
            if(users[i].id == id) {
                return i;
        }
        // if user does not exist then revert back
        revert("User does not exist");
    }
}
```

#### Practical No: 4

```
// SPDX-License-Identifier: MIT
pragma solidity ^0.8.0;
contract Crud {
    struct User {
        uint id;
        string name;
    }
    User[] public users;
    uint public nextId = 0;
    function Create(string memory name) public {
        users.push(User(nextId, name));
        nextId++;
    function Read(uint id) view public returns(uint, string memory) {
        for(uint i=0; i<users.length; i++) {</pre>
            if(users[i].id == id) {
                return(users[i].id, users[i].name);
            }
        }
    function Update(uint id, string memory name) public {
        for(uint i=0; i<users.length; i++) {</pre>
            if(users[i].id == id) {
                users[i].name =name;
        }
    function Delete(uint id) public {
        delete users[id];
    function find(uint id) view internal returns(uint) {
        for(uint i=0; i< users.length; i++) {</pre>
            if(users[i].id == id) {
                return i;
            }
        // if user does not exist then revert back
        revert("User does not exist");
    }
}
```