```
import pandas as pd
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
import time
# Load the Iris dataset
iris = load iris()
X = iris.data
y = iris.target
# Split the data 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)
# Initialize classifiers
models = {
    'Decision Tree': DecisionTreeClassifier(),
    'Logistic Regression': LogisticRegression(max_iter=1000),
    'KNN': KNeighborsClassifier()
}
# DataFrame to store results
results = pd.DataFrame(columns=['Algorithm', 'Accuracy', 'Training Time (s)', 'Prediction T
# Train and evaluate each model
for algo_name, model in models.items():
    # Measure training time
    start train = time.time()
    model.fit(X_train, y_train)
    end_train = time.time()
    train_time = end_train - start_train
    # Measure prediction time
    start pred = time.time()
    y_pred = model.predict(X_test)
    end_pred = time.time()
    pred_time = end_pred - start_pred
    # Calculate accuracy
    accuracy = accuracy_score(y_test, y_pred)
    # Create a DataFrame with the current results
    current_results = pd.DataFrame([{
        'Algorithm': algo_name,
        'Accuracy': accuracy,
        'Training Time (s)': train_time,
        'Prediction Time (s)': pred_time
    }])
```

```
# Concatenate with the main results DataFrame
results = pd.concat([results, current_results], ignore_index=True)
```

Display results
print("Performance Comparison of Classification Algorithms on Iris Dataset:\n")
print(results)

→ Performance Comparison of Classification Algorithms on Iris Dataset:

	Algorithm	Accuracy	Training Time (s)	Prediction Time (s)
0	Decision Tree	1.0	0.005356	0.000264
1	Logistic Regression	1.0	0.046542	0.000318
2	KNN	1.0	0.001233	0.004997

```
import numpy as np
def candidate elimination(examples):
    num_attributes = len(examples[0]) - 1 # Number of attributes in the examples
    specific_hypothesis = ['0'] * num_attributes # most specific hypothesis
    general_hypothesis = [['?'] * num_attributes] # most general hypothesis
    for example in examples:
        if example[-1] == 'Yes': # positive example
            for i in range(num_attributes):
                if specific hypothesis[i] == '0':
                    specific_hypothesis[i] = example[i]
                elif specific_hypothesis[i] != example[i]:
                    specific hypothesis[i] = '?'
            # Remove inconsistent hypotheses from general_hypothesis
            general_hypothesis = [g \text{ for } g \text{ in general_hypothesis if all}(g[i] == '?' \text{ or } g[i]
        else: # negative example
            new_general_hypotheses = []
            for g in general_hypothesis:
                for i in range(num_attributes):
                    if g[i] == '?':
                        if example[i] != specific_hypothesis[i]:
                            new_hypothesis = g[:]
                            new_hypothesis[i] = specific_hypothesis[i]
                            if new_hypothesis not in new_general_hypotheses:
                                new general hypotheses.append(new hypothesis)
            general_hypothesis = new_general_hypotheses
    return specific_hypothesis, general_hypothesis
# Example dataset
dataset = [
    ['Some', 'Small', 'No', 'Affordable', 'Few', 'No'],
    ['Many', 'Big', 'No', 'Expensive', 'Many', 'Yes'],
    ['Many', 'Medium', 'No', 'Expensive', 'Few', 'Yes'],
    ['Many', 'Small', 'No', 'Affordable', 'Many', 'Yes']
1
# Applying the Candidate-Elimination algorithm
specific, general = candidate_elimination(dataset)
print("Specific hypothesis:", specific)
print("General hypotheses:", general)
→ Specific hypothesis: ['Many', '?', 'No', '?', '?']
     General hypotheses: []
```

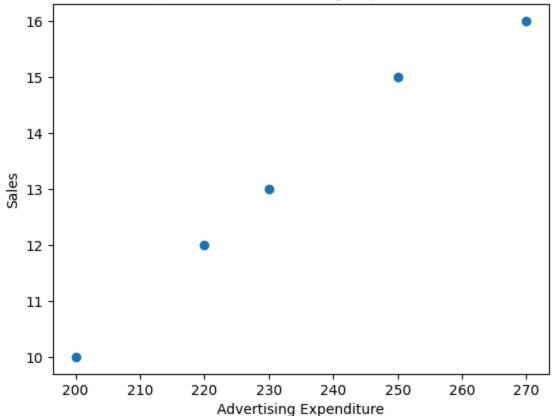
```
import pandas as pd
# Define the dataset
data = {
    'Example': [1, 2, 3, 4],
    'Citations': ['Some', 'Many', 'Many', 'Many'],
    'Size': ['Small', 'Big', 'Medium', 'Small'],
    'In Library': ['No', 'No', 'No', 'No'],
    'Price': ['Affordable', 'Expensive', 'Expensive', 'Affordable'],
    'Editions': ['Few', 'Many', 'Few', 'Many'],
    'Buy': ['No', 'Yes', 'Yes']
}
df = pd.DataFrame(data)
# Initialize the hypothesis
hypothesis = None
# Find-S algorithm
for index, row in df.iterrows():
    if row['Buy'] == 'Yes':
        if hypothesis is None:
            hypothesis = row.drop('Example')
        else:
            for attr in hypothesis.index:
                if hypothesis[attr] != row[attr]:
                    hypothesis[attr] = '?'
# Output the hypothesis
print("The most specific hypothesis is:")
print(hypothesis)
→ The most specific hypothesis is:
     Citations
                   Many
     Size
                      ?
     In Library
                     No
     Price
                      ?
     Editions
                      ?
     Buy
                    Yes
     Name: 1, dtype: object
```

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean squared error
import matplotlib.pyplot as plt
# Assuming 'data' is your dictionary with keys 'sales', 'advertising', and 'quarter'
data_dict = {
    'sales': [10, 12, 13, 15, 16],
    'advertising': [200, 220, 230, 250, 270],
    'quarter': [1, 2, 3, 4, 1]
}
# Convert dictionary to DataFrame
data = pd.DataFrame(data_dict)
# a) Print the first five rows
print(data.head())
# b) Basic statistical computations
print(data.describe())
# c) Columns and their data types
print(data.dtypes)
# d) Explore the data using scatter plot
plt.scatter(data['advertising'], data['sales'])
plt.xlabel('Advertising Expenditure')
plt.ylabel('Sales')
plt.title('Relationship between Advertising Expenditure and Sales')
plt.show()
# e) Handle null values by replacing with mode (if any)
data = data.fillna(data.mode().iloc[0])
# f) Split data into train and test sets (assuming 80-20 split)
X = data[['advertising']] # Feature
y = data['sales']
                           # Target
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# g) Train and predict using linear regression
model = LinearRegression()
model.fit(X_train, y_train)
# Predict on the test set
y_pred = model.predict(X_test)
# Print the coefficients and intercept
print("Coefficients:", model.coef_)
print("Intercept:", model.intercept_)
```

```
# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)
```

\rightarrow	sa	les	advert	ising	quarte	r	
	0	10		200		1	
	1	12		220		2	
	2	13		230		3	
	3	15		250		4	
	4	16		270		1	
			sales	s adve	ertising	5	quarter
	count	5.	.000000) 5	.000000)	5.00000
	mean	13.	200000	234	1.000000)	2.20000
	std	2.	387467	7 27	7.018512	2	1.30384
	min	10.	.000000	200	0.000000)	1.00000
	25%	12.	.000000	226	0.000000)	1.00000
	50%	13.	.000000	236	0.000000)	2.00000
	75%	15.	.000000	250	0.000000)	3.00000
	max	16.	.000000	276	0.000000)	4.00000
	sales		int64				
	advertisin		ng int64				
	quarter		j	int64			
	dtype	: obj	ject				

Relationship between Advertising Expenditure and Sales



Coefficients: [0.08785047] Intercept: -7.3644859813084125

Mean Squared Error: 0.001397501965237305

```
import pandas as pd

# Define the dataset
data = {
    'Example': [1, 2, 3, 4],
    'Shape': ['Circular', 'Circular', 'Oval', 'Oval'],
    'Size': ['Large', 'Large', 'Large', 'Large'],
    'Color': ['Light', 'Light', 'Dark', 'Light'],
    'Surface': ['Smooth', 'Irregular', 'Smooth', 'Irregular'],
    'Thickness': ['Thick', 'Thick', 'Thin', 'Thick'],
    'Target Concept': ['+', '+', '-', '+']
}

df = pd.DataFrame(data)

# Initialize the hypothesis
hypothesis = ['0', '0', '0', '0', '0']
```