


```
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Saving dataset.csv to dataset.csv

```
import pandas as pd
import numpy as np
```


```
df=pd.read_csv('dataset.csv')
```

df




	sky	temp	humidity	wind	isplay
0	sunny	warm	high	strong	yes
1	sunny	warm	normal	strong	yes
2	rainy	cold	high	strong	no
3	sunny	warm	high	less	yes

```
df.describe()
```




	sky	temp	humidity	wind	isplay
count	4	4	4	4	4
unique	2	2	2	2	2
top	sunny	warm	high	strong	yes
freq	3	3	3	3	3

```
df.info()
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4 entries, 0 to 3
Data columns (total 5 columns):
#   Column      Non-Null Count  Dtype
---  -
0    sky         4 non-null      object
1    temp        4 non-null      object
2    humidity    4 non-null      object
3    wind        4 non-null      object
4    isplay      4 non-null      object
dtypes: object(5)
memory usage: 292.0+ bytes
```

```
df.shape
```



```
(4, 5)
```

```
y=df["isplay"]
```

```
X=df.drop(columns=["isplay"])
```

X

```

sky temp humidity wind
0 sunny warm high strong
1 sunny warm normal strong
2 rainy cold high strong
3 sunny warm high less

```

y

```

isplay
0 yes
1 yes
2 no
3 yes

```

**dtype:** object

type(X)

```

pandas.core.frame.DataFrame
def __init__(data=None, index: Axes | None=None, columns: Axes | None=None, dtype: Dtype | None=None, copy: bool | None=None) -> None

Two-dimensional, size-mutable, potentially heterogeneous tabular data.

Data structure also contains labeled axes (rows and columns).
Arithmetic operations align on both row and column labels. Can be
thought of as a dict-like container for Series objects. The primary
pandas data structure.

```

type(y)

```

pandas.core.series.Series
def __init__(data=None, index=None, dtype: Dtype | None=None, name=None, copy: bool | None=None, fastpath: bool | lib.NoDefault=lib.no_default) -> None

One-dimensional ndarray with axis labels (including time series).

Labels need not be unique but must be a hashable type. The object
supports both integer- and label-based indexing and provides a host of
methods for performing operations involving the index. Statistical
methods from ndarray have been overridden to automatically exclude

```

X=np.array(X)

y=np.array(y)

```

def candidate_elimination(X, y):
    # Initialize the specific hypothesis (S) and general hypothesis (G)
    num_attributes = len(X[0])
    S = ['φ'] * num_attributes # Most specific hypothesis
    G = [['?'] * num_attributes] # Most general hypothesis

    # Iterate through the dataset
    for i, instance in enumerate(X):
        if y[i] == 'yes': # Positive example
            # Update S to be consistent with the instance
            for j in range(num_attributes):
                if S[j] == 'φ': # Replace 'φ' with the attribute value
                    S[j] = instance[j]
                elif S[j] != instance[j]: # Generalize to '?'
                    S[j] = '?'

            # Remove inconsistent hypotheses from G
            G = [g for g in G if all(
                g[k] == '?' or S[k] == '?' or g[k] == S[k]
                for k in range(num_attributes)
            )]

```

```

elif y[i] == 'no': # Negative example
    # Specialize G to exclude the negative instance
    new_G = []
    for g in G:
        for j in range(num_attributes):
            if g[j] == '?': # Specialize this attribute
                if S[j] != '?':
                    new_hypothesis = g.copy()
                    new_hypothesis[j] = S[j]
                    new_G.append(new_hypothesis)

    G = new_G

    # Remove inconsistent hypotheses from G
    G = [g for g in G if any(
        g[k] != instance[k] and g[k] != '?' for k in range(num_attributes)
    )]

return S, G

```

```

S_final, G_final = candidate_elimination(X, y)

```

```

# Output results

```

```

print("Final Specific Hypothesis:", S_final)

```

```

print("Final General Hypothesis:", G_final)

```

```

➡ Final Specific Hypothesis: ['sunny', 'warm', '?', '?']
  Final General Hypothesis: [['sunny', '?', '?', '?'], ['?', 'warm', '?', '?']]

```