

lab2_pandas

September 9, 2024

```
[11]: import pandas as pd
      print(pd.__version__)
```

2.2.2

```
[12]: # create a numeric series
list_location1 = [-121.87, 37.23, 19, 7357, 963, 3018, 981, 6.9473]
pd_location1 = pd.Series(list_location1)

print(pd_location1)

print("type(pd_location1): ", type(pd_location1))
```

```
0    -121.8700
1      37.2300
2      19.0000
3    7357.0000
4      963.0000
5    3018.0000
6      981.0000
7       6.9473
dtype: float64
type(pd_location1):  <class 'pandas.core.series.Series'>
```

```
[13]: #To create an object series (for column names)
table_column = ['Longitude', 'Latitude', 'Housing_median_age (Year)',
               ↪ 'Total_rooms', 'Total_bedrooms', 'Population', 'Households', 'Median_income']
pd_table_column = pd.Series(table_column)
print(pd_table_column)
```

```
0                Longitude
1                Latitude
2    Housing_median_age (Year)
3                Total_rooms
4            Total_bedrooms
5                Population
6                Households
7            Median_income
dtype: object
```

```
[14]: #To create a series by giving both numeric and string values
table_values = ['Longitude', 'Latitude', 25, 'Total_rooms', 'Total_bedrooms',
↪10.3, 'Households', 'Median_income']
pd_table_values = pd.Series(table_values)
print(pd_table_values)
```

```
0      Longitude
1      Latitude
2           25
3    Total_rooms
4  Total_bedrooms
5           10.3
6    Households
7    Median_income
dtype: object
```

```
[15]: # To create a series from a dictionary
longitude_column = {'location1': -113, 'location2': -111.2, 'location3': -112,
↪'location4': -110}
pd.Series(longitude_column)
```

```
[15]: location1    -113.0
location2    -111.2
location3    -112.0
location4    -110.0
dtype: float64
```

```
[16]: # Creating a 2d data frame a dictionary
data = {'Sample': ['location1', 'location2', 'location3', 'location4',
↪'location5', 'location6'],
'Longitude': [-121.87, -121.12, -119.31, -118.03, -120.97, -118.18],
'Latitude': [37.23, 39.03, 36.06, 33.78, 37.61, 34.02]}

df_data = pd.DataFrame(data)

print(df_data)
```

```
      Sample  Longitude  Latitude
0  location1    -121.87    37.23
1  location2    -121.12    39.03
2  location3    -119.31    36.06
3  location4    -118.03    33.78
4  location5    -120.97    37.61
5  location6    -118.18    34.02
```

0.1 Problem 1

Task 1

```
[17]: # Load the 'housing.csv' Download housing.csv' data into python using Pandas
      ↪ library.
      housing = pd.read_csv('housing-4-1.csv')
```

Task 2

```
[18]: print(housing.describe())
```

	longitude	latitude	housing_median_age	total_rooms \
count	20640.000000	20640.000000	20640.000000	20640.000000
mean	-119.569704	35.631861	28.639486	2635.763081
std	2.003532	2.135952	12.585558	2181.615252
min	-124.350000	32.540000	1.000000	2.000000
25%	-121.800000	33.930000	18.000000	1447.750000
50%	-118.490000	34.260000	29.000000	2127.000000
75%	-118.010000	37.710000	37.000000	3148.000000
max	-114.310000	41.950000	52.000000	39320.000000

	total_bedrooms	population	households	median_income \
count	20433.000000	20640.000000	20640.000000	20640.000000
mean	537.870553	1425.476744	499.539680	3.870671
std	421.385070	1132.462122	382.329753	1.899822
min	1.000000	3.000000	1.000000	0.499900
25%	296.000000	787.000000	280.000000	2.563400
50%	435.000000	1166.000000	409.000000	3.534800
75%	647.000000	1725.000000	605.000000	4.743250
max	6445.000000	35682.000000	6082.000000	15.000100

	median_house_value
count	20640.000000
mean	206855.816909
std	115395.615874
min	14999.000000
25%	119600.000000
50%	179700.000000
75%	264725.000000
max	500001.000000

Task 3

```
[19]: ###Check the Data types of values in each column of dataset using Pandas
      ↪ function (i.e., info()). Write code for the activity and print out the
      ↪ dataframe information.
      print(housing.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20640 entries, 0 to 20639
Data columns (total 9 columns):
 #   Column              Non-Null Count  Dtype
---  -
 #   Column              Non-Null Count  Dtype
```

```

0  longitude          20640 non-null float64
1  latitude           20640 non-null float64
2  housing_median_age 20640 non-null float64
3  total_rooms         20640 non-null float64
4  total_bedrooms      20433 non-null float64
5  population          20640 non-null float64
6  households          20640 non-null float64
7  median_income       20640 non-null float64
8  median_house_value  20640 non-null float64
dtypes: float64(9)
memory usage: 1.4 MB
None

```

Task 4

```
[21]: %pip install matplotlib
```

```

Collecting matplotlib
  Downloading matplotlib-3.9.2-cp311-cp311-win_amd64.whl.metadata (11 kB)
Collecting contourpy>=1.0.1 (from matplotlib)
  Downloading contourpy-1.3.0-cp311-cp311-win_amd64.whl.metadata (5.4 kB)
Collecting cyclor>=0.10 (from matplotlib)
  Downloading cyclor-0.12.1-py3-none-any.whl.metadata (3.8 kB)
Collecting fonttools>=4.22.0 (from matplotlib)
  Downloading fonttools-4.53.1-cp311-cp311-win_amd64.whl.metadata (165 kB)
----- 0.0/165.9 kB ? eta -:-:--
----- 143.4/165.9 kB 2.8 MB/s eta 0:00:01
----- 165.9/165.9 kB 3.3 MB/s eta 0:00:00
Collecting kiwisolver>=1.3.1 (from matplotlib)
  Downloading kiwisolver-1.4.7-cp311-cp311-win_amd64.whl.metadata (6.4 kB)
Requirement already satisfied: numpy>=1.23 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
matplotlib) (1.25.0)
Requirement already satisfied: packaging>=20.0 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
matplotlib) (23.2)
Requirement already satisfied: pillow>=8 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
matplotlib) (10.0.0)
Requirement already satisfied: pyparsing>=2.3.1 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
matplotlib) (3.1.0)
Requirement already satisfied: python-dateutil>=2.7 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
matplotlib) (2.8.2)
Requirement already satisfied: six>=1.5 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
python-dateutil>=2.7->matplotlib) (1.16.0)
Downloading matplotlib-3.9.2-cp311-cp311-win_amd64.whl (7.8 MB)

```

```

----- 0.0/7.8 MB ? eta -:--:--
-- ----- 0.6/7.8 MB 12.2 MB/s eta 0:00:01
----- 2.0/7.8 MB 20.9 MB/s eta 0:00:01
----- 4.3/7.8 MB 34.3 MB/s eta 0:00:01
----- 7.8/7.8 MB 45.7 MB/s eta 0:00:01
----- 7.8/7.8 MB 41.8 MB/s eta 0:00:00
Downloading contourpy-1.3.0-cp311-cp311-win_amd64.whl (217 kB)
----- 0.0/217.2 kB ? eta -:--:--
----- 217.2/217.2 kB 12.9 MB/s eta 0:00:00
Downloading cycler-0.12.1-py3-none-any.whl (8.3 kB)
Downloading fonttools-4.53.1-cp311-cp311-win_amd64.whl (2.2 MB)
----- 0.0/2.2 MB ? eta -:--:--
----- 2.2/2.2 MB 46.7 MB/s eta 0:00:00
Downloading kiwisolver-1.4.7-cp311-cp311-win_amd64.whl (56 kB)
----- 0.0/56.0 kB ? eta -:--:--
----- 56.0/56.0 kB 2.9 MB/s eta 0:00:00
Installing collected packages: kiwisolver, fonttools, cycler, contourpy,
matplotlib
Successfully installed contourpy-1.3.0 cycler-0.12.1 fonttools-4.53.1
kiwisolver-1.4.7 matplotlib-3.9.2
Note: you may need to restart the kernel to use updated packages.

```

[notice] A new release of pip is available: 24.1.2 -> 24.2

[notice] To update, run: python.exe -m pip install --upgrade pip

```

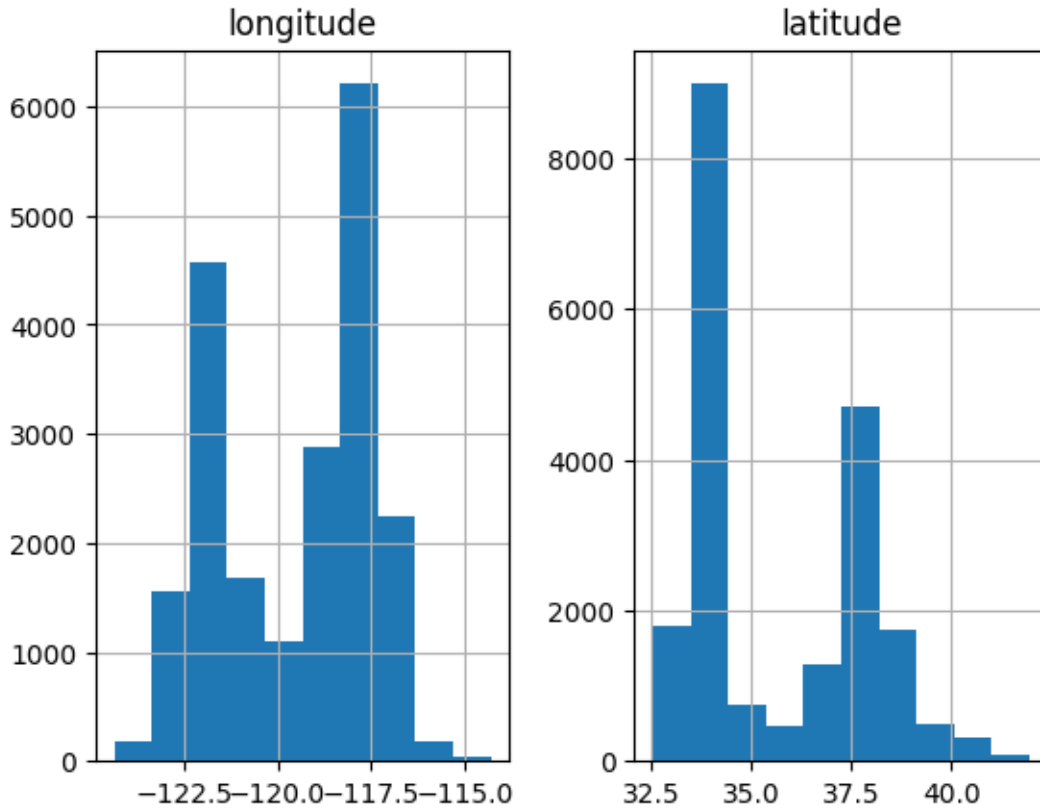
[22]: #Pick at least two numeric columns, and draw graph using pandas function (i.e.,
      ↪ histogram, scatter plot, etc.) Note: if no numeric columns are found in
      ↪ dataset, provide justification that none of columns in data are numeric
      ↪ columns
      pd.plotting.hist_frame(housing[['longitude', 'latitude']])

```

```

[22]: array([[<Axes: title={'center': 'longitude'}>,
              <Axes: title={'center': 'latitude'}>]], dtype=object)

```



The Kernel crashed while executing code in the current cell or a previous cell.

Please review the code in the cell(s) to identify a possible cause of the failure.

Click

View Jupyter

Task 5 There are no categorical columns because there is no qualitative data here, just quantitative data.

0.2 Problem 2

Task 1 This dataset, related to health and medicine, contains 32 instances and 56 integer-based features, with missing values. It is used for a classification task to distinguish between three types of pathological lung cancers. Models such as RDA, KNN, and the optimal discriminant plane have been applied, with varying accuracy rates (62.5%, 53.1%, and 59.4%, respectively). The data was used to illustrate the power of the optimal discriminant plane, despite being ill-posed, with some

features containing missing or altered values.

Task 2 i have no idea how to import the zip into python fso im doing it the other way listed on this website

```
[1]: %pip install ucimlrepo
```

```
Collecting ucimlrepo
  Downloading ucimlrepo-0.0.7-py3-none-any.whl.metadata (5.5 kB)
Requirement already satisfied: pandas>=1.0.0 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
ucimlrepo) (2.2.2)
Requirement already satisfied: certifi>=2020.12.5 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
ucimlrepo) (2024.7.4)
Requirement already satisfied: numpy>=1.23.2 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
pandas>=1.0.0->ucimlrepo) (1.25.0)
Requirement already satisfied: python-dateutil>=2.8.2 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
pandas>=1.0.0->ucimlrepo) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
pandas>=1.0.0->ucimlrepo) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
pandas>=1.0.0->ucimlrepo) (2023.3)
Requirement already satisfied: six>=1.5 in
c:\users\julian\appdata\local\programs\python\python311\lib\site-packages (from
python-dateutil>=2.8.2->pandas>=1.0.0->ucimlrepo) (1.16.0)
Downloading ucimlrepo-0.0.7-py3-none-any.whl (8.0 kB)
Installing collected packages: ucimlrepo
Successfully installed ucimlrepo-0.0.7
Note: you may need to restart the kernel to use updated packages.
```

```
[notice] A new release of pip is available: 24.1.2 -> 24.2
```

```
[notice] To update, run: python.exe -m pip install --upgrade pip
```

```
[4]: from ucimlrepo import fetch_ucirepo
lung_cancer = fetch_ucirepo(id=62)

print(lung_cancer)
```

```
{'data': {'ids': None, 'features':
Attribute1 Attribute2 Attribute3
Attribute4 Attribute5 Attribute6 \
0 0 3 0 NaN 0 2
1 0 3 3 1.0 0 3
2 0 3 3 2.0 0 3
```

3	0	2	3	2.0	1	3
4	0	3	2	1.0	1	3
5	0	3	3	2.0	0	3
6	0	3	2	1.0	0	3
7	0	2	2	1.0	0	3
8	0	3	1	1.0	0	3
9	0	2	3	2.0	0	2
10	0	2	2	0.0	0	3
11	0	2	3	2.0	0	1
12	0	2	1	1.0	0	1
13	0	2	2	1.0	1	2
14	1	3	0	NaN	1	1
15	0	3	2	2.0	1	2
16	0	3	2	2.0	0	1
17	0	2	1	1.0	0	2
18	0	2	0	NaN	0	2
19	0	1	2	1.0	0	3
20	0	2	0	NaN	1	3
21	0	3	3	2.0	0	2
22	0	2	3	1.0	1	2
23	0	2	3	1.0	1	1
24	0	3	3	1.0	0	3
25	0	2	3	2.0	0	1
26	0	2	2	2.0	0	2
27	0	2	2	1.0	0	2
28	0	3	2	2.0	0	2
29	0	2	1	1.0	0	2
30	0	2	3	2.0	1	2
31	0	2	3	1.0	0	2

	Attribute7	Attribute8	Attribute9	Attribute10	...	Attribute47	\
0	2	2	1	1	...	2	
1	1	3	1	1	...	2	
2	3	3	1	1	...	2	
3	3	3	1	2	...	2	
4	3	3	2	2	...	2	
5	3	3	1	2	...	2	
6	3	3	1	2	...	2	
7	1	3	3	3	...	2	
8	1	3	1	1	...	2	
9	2	2	1	2	...	2	
10	2	3	1	1	...	2	
11	2	1	1	2	...	2	
12	2	2	1	2	...	2	
13	3	3	1	1	...	2	
14	2	2	1	1	...	2	
15	2	2	1	1	...	2	
16	1	3	1	1	...	2	

17	1	3	1	1 ...	2
18	3	3	3	2 ...	2
19	3	3	1	2 ...	2
20	3	3	1	2 ...	2
21	1	3	1	1 ...	2
22	2	1	1	1 ...	3
23	2	1	1	1 ...	2
24	3	1	1	1 ...	2
25	2	2	1	2 ...	2
26	1	2	1	1 ...	2
27	2	2	1	1 ...	3
28	2	2	1	1 ...	2
29	2	1	1	1 ...	2
30	2	3	1	1 ...	2
31	3	3	1	1 ...	2

	Attribute48	Attribute49	Attribute50	Attribute51	Attribute52	\
0	2	2	2	2	1	
1	2	2	2	2	2	
2	2	2	2	2	2	
3	2	2	2	2	2	
4	2	2	2	2	2	
5	2	2	2	2	2	
6	2	2	2	1	2	
7	2	1	2	2	2	
8	2	2	2	2	2	
9	2	2	1	3	2	
10	2	2	2	2	2	
11	2	2	2	2	1	
12	2	2	2	2	2	
13	2	2	2	2	1	
14	2	2	2	2	2	
15	2	2	2	2	2	
16	2	2	2	2	2	
17	2	2	2	2	1	
18	2	2	2	2	2	
19	2	2	2	2	1	
20	2	2	2	1	2	
21	2	1	2	2	2	
22	3	3	3	1	3	
23	2	2	2	2	2	
24	2	2	2	3	2	
25	2	2	1	3	1	
26	2	2	2	2	2	
27	3	2	2	3	2	
28	2	2	3	1	2	
29	2	3	2	2	2	
30	2	2	2	2	2	

31	2	2	2	2	2
----	---	---	---	---	---

	Attribute53	Attribute54	Attribute55	Attribute56
0	1	1	2	2
1	2	1	2	2
2	2	2	1	2
3	2	2	2	2
4	2	1	2	2
5	2	2	1	2
6	2	2	1	2
7	2	1	2	2
8	2	1	2	2
9	1	1	2	2
10	2	2	2	2
11	1	2	2	1
12	2	1	2	2
13	1	1	2	2
14	2	1	2	1
15	2	2	2	2
16	2	1	2	2
17	1	1	2	2
18	2	2	1	2
19	1	2	2	1
20	2	1	2	2
21	2	2	1	2
22	3	2	2	1
23	2	2	2	1
24	2	2	2	1
25	2	2	1	2
26	2	1	2	1
27	2	2	2	1
28	2	2	2	2
29	2	2	2	1
30	2	1	2	2
31	2	2	2	2

[32 rows x 56 columns], 'targets': class

0	1
1	1
2	1
3	1
4	1
5	1
6	1
7	1
8	1
9	2
10	2

11	2					
12	2					
13	2					
14	2					
15	2					
16	2					
17	2					
18	2					
19	2					
20	2					
21	2					
22	3					
23	3					
24	3					
25	3					
26	3					
27	3					
28	3					
29	3					
30	3					
31	3, 'original':	class	Attribute1	Attribute2	Attribute3	Attribute4
Attribute5	\					
0	1	0	3	0	NaN	0
1	1	0	3	3	1.0	0
2	1	0	3	3	2.0	0
3	1	0	2	3	2.0	1
4	1	0	3	2	1.0	1
5	1	0	3	3	2.0	0
6	1	0	3	2	1.0	0
7	1	0	2	2	1.0	0
8	1	0	3	1	1.0	0
9	2	0	2	3	2.0	0
10	2	0	2	2	0.0	0
11	2	0	2	3	2.0	0
12	2	0	2	1	1.0	0
13	2	0	2	2	1.0	1
14	2	1	3	0	NaN	1
15	2	0	3	2	2.0	1
16	2	0	3	2	2.0	0
17	2	0	2	1	1.0	0
18	2	0	2	0	NaN	0
19	2	0	1	2	1.0	0
20	2	0	2	0	NaN	1
21	2	0	3	3	2.0	0
22	3	0	2	3	1.0	1
23	3	0	2	3	1.0	1
24	3	0	3	3	1.0	0
25	3	0	2	3	2.0	0

26	3	0	2	2	2.0	0
27	3	0	2	2	1.0	0
28	3	0	3	2	2.0	0
29	3	0	2	1	1.0	0
30	3	0	2	3	2.0	1
31	3	0	2	3	1.0	0

	Attribute6	Attribute7	Attribute8	Attribute9	...	Attribute47	\
0	2	2	2	1	...	2	
1	3	1	3	1	...	2	
2	3	3	3	1	...	2	
3	3	3	3	1	...	2	
4	3	3	3	2	...	2	
5	3	3	3	1	...	2	
6	3	3	3	1	...	2	
7	3	1	3	3	...	2	
8	3	1	3	1	...	2	
9	2	2	2	1	...	2	
10	3	2	3	1	...	2	
11	1	2	1	1	...	2	
12	1	2	2	1	...	2	
13	2	3	3	1	...	2	
14	1	2	2	1	...	2	
15	2	2	2	1	...	2	
16	1	1	3	1	...	2	
17	2	1	3	1	...	2	
18	2	3	3	3	...	2	
19	3	3	3	1	...	2	
20	3	3	3	1	...	2	
21	2	1	3	1	...	2	
22	2	2	1	1	...	3	
23	1	2	1	1	...	2	
24	3	3	1	1	...	2	
25	1	2	2	1	...	2	
26	2	1	2	1	...	2	
27	2	2	2	1	...	3	
28	2	2	2	1	...	2	
29	2	2	1	1	...	2	
30	2	2	3	1	...	2	
31	2	3	3	1	...	2	

	Attribute48	Attribute49	Attribute50	Attribute51	Attribute52	\
0	2	2	2	2	1	
1	2	2	2	2	2	
2	2	2	2	2	2	
3	2	2	2	2	2	
4	2	2	2	2	2	
5	2	2	2	2	2	

6	2	2	2	1	2
7	2	1	2	2	2
8	2	2	2	2	2
9	2	2	1	3	2
10	2	2	2	2	2
11	2	2	2	2	1
12	2	2	2	2	2
13	2	2	2	2	1
14	2	2	2	2	2
15	2	2	2	2	2
16	2	2	2	2	2
17	2	2	2	2	1
18	2	2	2	2	2
19	2	2	2	2	1
20	2	2	2	1	2
21	2	1	2	2	2
22	3	3	3	1	3
23	2	2	2	2	2
24	2	2	2	3	2
25	2	2	1	3	1
26	2	2	2	2	2
27	3	2	2	3	2
28	2	2	3	1	2
29	2	3	2	2	2
30	2	2	2	2	2
31	2	2	2	2	2

	Attribute53	Attribute54	Attribute55	Attribute56
0	1	1	2	2
1	2	1	2	2
2	2	2	1	2
3	2	2	2	2
4	2	1	2	2
5	2	2	1	2
6	2	2	1	2
7	2	1	2	2
8	2	1	2	2
9	1	1	2	2
10	2	2	2	2
11	1	2	2	1
12	2	1	2	2
13	1	1	2	2
14	2	1	2	1
15	2	2	2	2
16	2	1	2	2
17	1	1	2	2
18	2	2	1	2
19	1	2	2	1

20	2	1	2	2
21	2	2	1	2
22	3	2	2	1
23	2	2	2	1
24	2	2	2	1
25	2	2	1	2
26	2	1	2	1
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28	2	2	2	2
29	2	2	2	1
30	2	1	2	2
31	2	2	2	2

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'Attribute3', 'Attribute4',
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'Attribute10', 'Attribute11', 'Attribute12', 'Attribute13',
'Attribute14', 'Attribute15', 'Attribute16', 'Attribute17',
'Attribute18', 'Attribute19', 'Attribute20', 'Attribute21',
'Attribute22', 'Attribute23', 'Attribute24', 'Attribute25',
'Attribute26', 'Attribute27', 'Attribute28', 'Attribute29',
'Attribute30', 'Attribute31', 'Attribute32', 'Attribute33',
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'Attribute38', 'Attribute39', 'Attribute40', 'Attribute41',
'Attribute42', 'Attribute43', 'Attribute44', 'Attribute45',
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'dataset_doi': '10.24432/C57596', 'creators': ['Z.Q. Hong', 'J.Y Yang'],
'intro_paper': None, 'additional_info': {'summary': "This data was used by Hong
and Young to illustrate the power of the optimal discriminant plane even in ill-
posed settings. Applying the KNN method in the resulting plane gave 77%
accuracy. However, these results are strongly biased (See Aeberhard's second
ref. above, or email to stefan@coral.cs.jcu.edu.au). Results obtained by
Aeberhard et al. are :\r\n\r\nRDA : 62.5%, KNN 53.1%, Opt. Disc. Plane
59.4%\r\n\r\nThe data described 3 types of pathological lung cancers. The
Authors give no information on the individual variables nor on where the data
was originally used.\r\n\r\nNotes:\r\n- In the original data 4 values for the
fifth attribute were -1. These values have been changed to ? (unknown). (*)\r\n-
```

In the original data 1 value for the 39 attribute was 4. This value has been changed to ? (unknown). (*)", 'purpose': None, 'funded_by': None, 'instances_represent': None, 'recommended_data_splits': None, 'sensitive_data': None, 'preprocessing_description': None, 'variable_info': 'Attribute 1 is the class label.\r\n\r\nAll predictive attributes are nominal, taking on integer values 0-3', 'citation': None}}, 'variables':

					name	role
type	demographic	description	units	\		
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3	Attribute3	Feature	Categorical	None	None	None
4	Attribute4	Feature	Categorical	None	None	None
5	Attribute5	Feature	Categorical	None	None	None
6	Attribute6	Feature	Categorical	None	None	None
7	Attribute7	Feature	Categorical	None	None	None
8	Attribute8	Feature	Categorical	None	None	None
9	Attribute9	Feature	Categorical	None	None	None
10	Attribute10	Feature	Categorical	None	None	None
11	Attribute11	Feature	Categorical	None	None	None
12	Attribute12	Feature	Categorical	None	None	None
13	Attribute13	Feature	Categorical	None	None	None
14	Attribute14	Feature	Categorical	None	None	None
15	Attribute15	Feature	Categorical	None	None	None
16	Attribute16	Feature	Categorical	None	None	None
17	Attribute17	Feature	Categorical	None	None	None
18	Attribute18	Feature	Categorical	None	None	None
19	Attribute19	Feature	Categorical	None	None	None
20	Attribute20	Feature	Categorical	None	None	None
21	Attribute21	Feature	Categorical	None	None	None
22	Attribute22	Feature	Categorical	None	None	None
23	Attribute23	Feature	Categorical	None	None	None
24	Attribute24	Feature	Categorical	None	None	None
25	Attribute25	Feature	Categorical	None	None	None
26	Attribute26	Feature	Categorical	None	None	None
27	Attribute27	Feature	Categorical	None	None	None
28	Attribute28	Feature	Categorical	None	None	None
29	Attribute29	Feature	Categorical	None	None	None
30	Attribute30	Feature	Categorical	None	None	None
31	Attribute31	Feature	Categorical	None	None	None
32	Attribute32	Feature	Categorical	None	None	None
33	Attribute33	Feature	Categorical	None	None	None
34	Attribute34	Feature	Categorical	None	None	None
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37	Attribute37	Feature	Categorical	None	None	None
38	Attribute38	Feature	Categorical	None	None	None
39	Attribute39	Feature	Categorical	None	None	None
40	Attribute40	Feature	Categorical	None	None	None

41	Attribute41	Feature	Categorical	None	None	None
42	Attribute42	Feature	Categorical	None	None	None
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44	Attribute44	Feature	Categorical	None	None	None
45	Attribute45	Feature	Categorical	None	None	None
46	Attribute46	Feature	Categorical	None	None	None
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56	Attribute56	Feature	Categorical	None	None	None

missing_values	
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3	no
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41	no
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