

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

Task 1

```
df = pd.read_csv('insurance.csv', index_col=None)
df
```

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.92400
1	18	male	33.770	1	no	southeast	1725.55230
2	28	male	33.000	3	no	southeast	4449.46200
3	33	male	22.705	0	no	northwest	21984.47061
4	32	male	28.880	0	no	northwest	3866.85520
...
995	39	female	23.275	3	no	northeast	7986.47525
996	39	female	34.100	3	no	southwest	7418.52200
997	63	female	36.850	0	no	southeast	13887.96850
998	33	female	36.290	3	no	northeast	6551.75010
999	36	female	26.885	0	no	northwest	5267.81815

```
[1000 rows x 7 columns]
```

```
df.info
```

```
<bound method DataFrame.info of
smoker      region      charges
0      19  female  27.900      0    yes  southwest  16884.92400
1      18   male  33.770      1     no  southeast   1725.55230
2      28   male  33.000      3     no  southeast   4449.46200
3      33   male  22.705      0     no  northwest  21984.47061
4      32   male  28.880      0     no  northwest   3866.85520
...     ...     ...     ...     ...     ...     ...
995     39  female  23.275      3     no  northeast   7986.47525
996     39  female  34.100      3     no  southwest   7418.52200
997     63  female  36.850      0     no  southeast  13887.96850
998     33  female  36.290      3     no  northeast   6551.75010
999     36  female  26.885      0     no  northwest   5267.81815
```

```
[1000 rows x 7 columns]>
```

```
from sklearn.preprocessing import LabelEncoder
```

```
LE = LabelEncoder()
df['sex'] = LE.fit_transform(df['sex'])
df['smoker'] = LE.fit_transform(df['smoker'])
df['region'] = LE.fit_transform(df['region'])
```

```
df
```

	age	sex	bmi	children	smoker	region	charges
0	19	0	27.900	0	1	3	16884.92400
1	18	1	33.770	1	0	2	1725.55230
2	28	1	33.000	3	0	2	4449.46200
3	33	1	22.705	0	0	1	21984.47061
4	32	1	28.880	0	0	1	3866.85520
...
995	39	0	23.275	3	0	0	7986.47525
996	39	0	34.100	3	0	3	7418.52200
997	63	0	36.850	0	0	2	13887.96850
998	33	0	36.290	3	0	0	6551.75010
999	36	0	26.885	0	0	1	5267.81815

[1000 rows x 7 columns]

Label Encoding Info:

sex: Male - 1, Female - 0 smoker: Yes - 1, No - 0 region: southwest - 3, southeast - 2, northwest - 1, northeast - 0

```
data = df.iloc[:, :6].values
data
```

```
array([[19.    ,  0.    , 27.9   ,  0.    ,  1.    ,  3.    ],
       [18.    ,  1.    , 33.77  ,  1.    ,  0.    ,  2.    ],
       [28.    ,  1.    , 33.    ,  3.    ,  0.    ,  2.    ],
       ...,
       [63.    ,  0.    , 36.85  ,  0.    ,  0.    ,  2.    ],
       [33.    ,  0.    , 36.29  ,  3.    ,  0.    ,  0.    ],
       [36.    ,  0.    , 26.885 ,  0.    ,  0.    ,  1.    ]])
```

```
charges = df.iloc[:, -1].values
charges
```

```
array([16884.924 , 1725.5523 , 4449.462 , 21984.47061 ,
       3866.8552 , 3756.6216 , 8240.5896 , 7281.5056 ,
       6406.4107 , 28923.13692 , 2721.3208 , 27808.7251 ,
       1826.843 , 11090.7178 , 39611.7577 , 1837.237 ,
       10797.3362 , 2395.17155 , 10602.385 , 36837.467 ,
       13228.84695 , 4149.736 , 1137.011 , 37701.8768 ,
       6203.90175 , 14001.1338 , 14451.83515 , 12268.63225 ,
       2775.19215 , 38711. , 35585.576 , 2198.18985 ,
       4687.797 , 13770.0979 , 51194.55914 , 1625.43375 ,
       15612.19335 , 2302.3 , 39774.2763 , 48173.361 ,
       3046.062 , 4949.7587 , 6272.4772 , 6313.759 ,
       6079.6715 , 20630.28351 , 3393.35635 , 3556.9223 ,
       12629.8967 , 38709.176 , 2211.13075 , 3579.8287 ,
       23568.272 , 37742.5757 , 8059.6791 , 47496.49445 ,
       13607.36875 , 34303.1672 , 23244.7902 , 5989.52365 ,
       8606.2174 , 4504.6624 , 30166.61817 , 4133.64165 ,
```

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3947.4131	,	1532.4697	,	2755.02095	,	6571.02435	,
4441.21315	,	7935.29115	,	37165.1638	,	11033.6617	,
39836.519	,	21098.55405	,	43578.9394	,	11073.176	,
8026.6666	,	11082.5772	,	2026.9741	,	10942.13205	,
30184.9367	,	5729.0053	,	47291.055	,	3766.8838	,
12105.32	,	10226.2842	,	22412.6485	,	15820.699	,
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2867.1196	,	47055.5321	,	10825.2537	,	11881.358	,
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11381.3254	,	19107.7796	,	8601.3293	,	6686.4313	,
7740.337	,	1705.6245	,	2257.47525	,	39556.4945	,
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1261.442	,	2045.68525	,	27322.73386	,	2166.732	,
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20745.9891	,	5138.2567	,	40720.55105	,	9877.6077	,
10959.6947	,	1842.519	,	5125.2157	,	7789.635	,
6334.34355	,	19964.7463	,	7077.1894	,	6948.7008	,
21223.6758	,	15518.18025	,	36950.2567	,	19749.38338	,
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5028.1466	,	10407.08585	,	4830.63	,	6128.79745	,
2719.27975	,	4827.90495	,	13405.3903	,	8116.68	,
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6455.86265	,	10436.096	,	8823.279	,	8538.28845	,
11735.87905	,	1631.8212	,	4005.4225	,	7419.4779	,
7731.4271	,	43753.33705	,	3981.9768	,	5325.651	,
6775.961	,	4922.9159	,	12557.6053	,	4883.866	,
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4435.0942	,	39241.442	,	8547.6913	,	6571.544	,
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2731.9122	,	21195.818	,	7209.4918	,	18310.742	,
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7046.7222	,	14313.8463	,	2103.08	,	38792.6856	,
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7445.918	,	2680.9493	,	1621.8827	,	8219.2039	,
12523.6048	,	16069.08475	,	43813.8661	,	20773.62775	,
39597.4072	,	6117.4945	,	13393.756	,	5266.3656	,
4719.73655	,	11743.9341	,	5377.4578	,	7160.3303	,
4402.233	,	11657.7189	,	6402.29135	,	12622.1795	,
1526.312	,	12323.936	,	36021.0112	,	27533.9129	,
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13129.60345 , 4391.652 , 8457.818 , 3392.3652 ,
5966.8874 , 6849.026 , 8891.1395 , 2690.1138 ,
26140.3603 , 6653.7886 , 6282.235 , 6311.952 ,
3443.064 , 2789.0574 , 2585.85065 , 46255.1125 ,
4877.98105 , 19719.6947 , 27218.43725 , 5272.1758 ,
1682.597 , 11945.1327 , 29330.98315 , 7243.8136 ,
10422.91665 , 44202.6536 , 13555.0049 , 13063.883 ,
19798.05455 , 2221.56445 , 1634.5734 , 2117.33885 ,
8688.85885 , 48673.5588 , 4661.28635 , 8125.7845 ,
12644.589 , 4564.19145 , 4846.92015 , 7633.7206 ,
15170.069 , 17496.306 , 2639.0429 , 33732.6867 ,
14382.70905 , 7626.993 , 5257.50795 , 2473.3341 ,
21774.32215 , 35069.37452 , 13041.921 , 5245.2269 ,
13451.122 , 13462.52 , 5488.262 , 4320.41085 ,
6250.435 , 25333.33284 , 2913.569 , 12032.326 ,
13470.8044 , 6289.7549 , 2927.0647 , 6238.298 ,
10096.97 , 7348.142 , 4673.3922 , 12233.828 ,
32108.66282 , 8965.79575 , 2304.0022 , 9487.6442 ,
1121.8739 , 9549.5651 , 2217.46915 , 1628.4709 ,
12982.8747 , 11674.13 , 7160.094 , 39047.285 ,
6358.77645 , 19933.458 , 11534.87265 , 47462.894 ,
4527.18295 , 38998.546 , 20009.63365 , 3875.7341 ,
41999.52 , 12609.88702 , 41034.2214 , 28468.91901 ,
2730.10785 , 3353.284 , 14474.675 , 9500.57305 ,
26467.09737 , 4746.344 , 23967.38305 , 7518.02535 ,
3279.86855 , 8596.8278 , 10702.6424 , 4992.3764 ,
2527.81865 , 1759.338 , 2322.6218 , 16138.76205 ,
7804.1605 , 2902.9065 , 9704.66805 , 4889.0368 ,
25517.11363 , 4500.33925 , 19199.944 , 16796.41194 ,
4915.05985 , 7624.63 , 8410.04685 , 28340.18885 ,
4518.82625 , 14571.8908 , 3378.91 , 7144.86265 ,
10118.424 , 5484.4673 , 16420.49455 , 7986.47525 ,
7418.522 , 13887.9685 , 6551.7501 , 5267.81815 ] )

```

```

from sklearn import linear_model
from sklearn.metrics import mean_squared_error, r2_score

regr = linear_model.LinearRegression()

regr.fit(data, charges)

y_pred = regr.predict(data)

print("Coefficients: \n", regr.coef_)
print("Mean squared error: %.2f" % mean_squared_error(charges,

```


4	M1230701703	0.0	0.0	0
0				
...
...				
1047295	M1938858197	0.0	0.0	0
0				
1047296	M391493091	0.0	0.0	0
0				
1047297	M227905058	0.0	0.0	0
0				
1047298	M1154565434	0.0	0.0	0
0				
1047299	M919485291	0.0	0.0	0
0				

[1047300 rows x 11 columns]

df.info

```
<bound method DataFrame.info of
nameOrig  oldbalanceOrig  newbalanceOrig  step  type  amount
0          1  PAYMENT    9839.64  C1231006815  170136.00
160296.36
1          1  PAYMENT    1864.28  C1666544295   21249.00
19384.72
2          1  TRANSFER    181.00  C1305486145    181.00
0.00
3          1  CASH_OUT    181.00  C840083671    181.00
0.00
4          1  PAYMENT   11668.14  C2048537720   41554.00
29885.86
...      ...      ...      ...      ...      ...
...
1047295    95  PAYMENT    9028.31  C351212668   105818.00
96789.69
1047296    95  PAYMENT    6290.51  C404585795    96789.69
90499.18
1047297    95  PAYMENT    2499.92  C2079661953    90499.18
87999.25
1047298    95  PAYMENT    4780.32  C2115046838     802.00
0.00
1047299    95  PAYMENT     789.77  C386614070   30791.00
30001.23
```

	nameDest	oldbalanceDest	newbalanceDest	isFraud
isFlaggedFraud				
0	M1979787155	0.0	0.0	0
0				
1	M2044282225	0.0	0.0	0
0				

2	C553264065	0.0	0.0	1
0				
3	C38997010	21182.0	0.0	1
0				
4	M1230701703	0.0	0.0	0
0				
...
...				
1047295	M1938858197	0.0	0.0	0
0				
1047296	M391493091	0.0	0.0	0
0				
1047297	M227905058	0.0	0.0	0
0				
1047298	M1154565434	0.0	0.0	0
0				
1047299	M919485291	0.0	0.0	0
0				

[1047300 rows x 11 columns]>

```
from sklearn.preprocessing import LabelEncoder
```

```
LE = LabelEncoder()
df['type'] = LE.fit_transform(df['type'])
df['nameOrig'] = LE.fit_transform(df['nameOrig'])
df['nameDest'] = LE.fit_transform(df['nameDest'])
```

df

	step	type	amount	nameOrig	oldbalanceOrg	newbalanceOrig
\						
0	1	3	9839.64	125004	170136.00	160296.36
1	1	3	1864.28	360882	21249.00	19384.72
2	1	4	181.00	165023	181.00	0.00
3	1	1	181.00	960498	181.00	0.00
4	1	3	11668.14	567221	41554.00	29885.86
...
1047295	95	3	9028.31	694724	105818.00	96789.69
1047296	95	3	6290.51	723239	96789.69	90499.18
1047297	95	3	2499.92	583942	90499.18	87999.25
1047298	95	3	4780.32	603263	802.00	0.00

1047299	95	3	789.77	713704	30791.00	30001.23
---------	----	---	--------	--------	----------	----------

	nameDest	oldbalanceDest	newbalanceDest	isFraud
isFlaggedFraud				
0	274496	0.0	0.0	0
0				
1	286268	0.0	0.0	0
0				
2	73503	0.0	0.0	1
0				
3	65423	21182.0	0.0	1
0				
4	137860	0.0	0.0	0
0				
...
...				
1047295	266884	0.0	0.0	0
0				
1047296	337296	0.0	0.0	0
0				
1047297	307566	0.0	0.0	0
0				
1047298	124109	0.0	0.0	0
0				
1047299	433968	0.0	0.0	0
0				

[1047300 rows x 11 columns]

```
data = df.iloc[:,10].values
data
```

```
array([[1.00000e+00, 3.00000e+00, 9.83964e+03, ..., 0.00000e+00,
        0.00000e+00, 0.00000e+00],
       [1.00000e+00, 3.00000e+00, 1.86428e+03, ..., 0.00000e+00,
        0.00000e+00, 0.00000e+00],
       [1.00000e+00, 4.00000e+00, 1.81000e+02, ..., 0.00000e+00,
        0.00000e+00, 1.00000e+00],
       ...,
       [9.50000e+01, 3.00000e+00, 2.49992e+03, ..., 0.00000e+00,
        0.00000e+00, 0.00000e+00],
       [9.50000e+01, 3.00000e+00, 4.78032e+03, ..., 0.00000e+00,
        0.00000e+00, 0.00000e+00],
       [9.50000e+01, 3.00000e+00, 7.89770e+02, ..., 0.00000e+00,
        0.00000e+00, 0.00000e+00]])
```

```
label = df.iloc[:,2].values
label
```

```

array([0, 0, 1, ..., 0, 0, 0])

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(data, label,
test_size=0.2, random_state=42)

from sklearn.linear_model import LogisticRegression

logreg = LogisticRegression(random_state=16)

logreg.fit(X_train, y_train)

LogisticRegression(random_state=16)

y_pred = logreg.predict(X_test)

from sklearn import metrics

cm = metrics.confusion_matrix(y_test, y_pred)
cm

array([[209217,    22],
       [   138,    83]])

from sklearn.metrics import classification_report
target_names = ['not spam', 'spam']
print(classification_report(y_test, y_pred,
target_names=target_names))

```

	precision	recall	f1-score	support
not spam	1.00	1.00	1.00	209239
spam	0.79	0.38	0.51	221
accuracy			1.00	209460
macro avg	0.89	0.69	0.75	209460
weighted avg	1.00	1.00	1.00	209460

```

acc = metrics.accuracy_score(y_test, y_pred)
print("Accuracy: "+str(acc))

Accuracy: 0.9992361310035329

```