

## Initial data exploration

### Load

```
import pandas
```

```
data = pandas.read_csv('../data/weatherAUS.csv')
```

### Data shape and head

Notes:

- NaN are missing values.
- Response variable to predict: RainTomorrow (Yes, No)

```
print('Dimensions of data: {}'.format(data.shape))
print('Head:')
data.head(n=10)
```

Dimensions of data: (145460, 23)

Head:

	Date	Location	MinTemp	MaxTemp	Rainfall	Evaporation
Sunshine \						
0	2008-12-01	Albury	13.4	22.9	0.6	NaN
NaN						
1	2008-12-02	Albury	7.4	25.1	0.0	NaN
NaN						
2	2008-12-03	Albury	12.9	25.7	0.0	NaN
NaN						
3	2008-12-04	Albury	9.2	28.0	0.0	NaN
NaN						
4	2008-12-05	Albury	17.5	32.3	1.0	NaN
NaN						
5	2008-12-06	Albury	14.6	29.7	0.2	NaN
NaN						
6	2008-12-07	Albury	14.3	25.0	0.0	NaN
NaN						
7	2008-12-08	Albury	7.7	26.7	0.0	NaN
NaN						
8	2008-12-09	Albury	9.7	31.9	0.0	NaN
NaN						
9	2008-12-10	Albury	13.1	30.1	1.4	NaN
NaN						

	WindGustDir	WindGustSpeed	WindDir9am	...	Humidity9am \
0	W	44.0	W	...	71.0
1	WNW	44.0	NNW	...	44.0
2	WSW	46.0	W	...	38.0

3	NE	24.0	SE	...	45.0
4	W	41.0	ENE	...	82.0
5	WNW	56.0	W	...	55.0
6	W	50.0	SW	...	49.0
7	W	35.0	SSE	...	48.0
8	NNW	80.0	SE	...	42.0
9	W	28.0	S	...	58.0

	Humidity3pm	Pressure9am	Pressure3pm	Cloud9am	Cloud3pm	Temp9am
\						
0	22.0	1007.7	1007.1	8.0	NaN	16.9
1	25.0	1010.6	1007.8	NaN	NaN	17.2
2	30.0	1007.6	1008.7	NaN	2.0	21.0
3	16.0	1017.6	1012.8	NaN	NaN	18.1
4	33.0	1010.8	1006.0	7.0	8.0	17.8
5	23.0	1009.2	1005.4	NaN	NaN	20.6
6	19.0	1009.6	1008.2	1.0	NaN	18.1
7	19.0	1013.4	1010.1	NaN	NaN	16.3
8	9.0	1008.9	1003.6	NaN	NaN	18.3
9	27.0	1007.0	1005.7	NaN	NaN	20.1

	Temp3pm	RainToday	RainTomorrow
0	21.8	No	No
1	24.3	No	No
2	23.2	No	No
3	26.5	No	No
4	29.7	No	No
5	28.9	No	No
6	24.6	No	No
7	25.5	No	No
8	30.2	No	Yes
9	28.2	Yes	No

[10 rows x 23 columns]

## Data summaries

Numerical variables:

```
print('Total number of rows: {0}'.format(data.shape[0]))
data.describe()
```

Total number of rows: 145460

	MinTemp	MaxTemp	Rainfall	Evaporation	\
count	143975.000000	144199.000000	142199.000000	82670.000000	
mean	12.194034	23.221348	2.360918	5.468232	
std	6.398495	7.119049	8.478060	4.193704	
min	-8.500000	-4.800000	0.000000	0.000000	
25%	7.600000	17.900000	0.000000	2.600000	
50%	12.000000	22.600000	0.000000	4.800000	
75%	16.900000	28.200000	0.800000	7.400000	
max	33.900000	48.100000	371.000000	145.000000	

	Sunshine	WindGustSpeed	WindSpeed9am	WindSpeed3pm	\
count	75625.000000	135197.000000	143693.000000	142398.000000	
mean	7.611178	40.035230	14.043426	18.662657	
std	3.785483	13.607062	8.915375	8.809800	
min	0.000000	6.000000	0.000000	0.000000	
25%	4.800000	31.000000	7.000000	13.000000	
50%	8.400000	39.000000	13.000000	19.000000	
75%	10.600000	48.000000	19.000000	24.000000	
max	14.500000	135.000000	130.000000	87.000000	

	Humidity9am	Humidity3pm	Pressure9am	Pressure3pm	\
count	142806.000000	140953.000000	130395.000000	130432.000000	
mean	68.880831	51.539116	1017.64994	1015.255889	
std	19.029164	20.795902	7.10653	7.037414	
min	0.000000	0.000000	980.50000	977.100000	
25%	57.000000	37.000000	1012.90000	1010.400000	
50%	70.000000	52.000000	1017.60000	1015.200000	
75%	83.000000	66.000000	1022.40000	1020.000000	
max	100.000000	100.000000	1041.00000	1039.600000	

	Cloud9am	Cloud3pm	Temp9am	Temp3pm
count	89572.000000	86102.000000	143693.000000	141851.000000
mean	4.447461	4.509930	16.990631	21.68339
std	2.887159	2.720357	6.488753	6.93665
min	0.000000	0.000000	-7.200000	-5.40000
25%	1.000000	2.000000	12.300000	16.60000
50%	5.000000	5.000000	16.700000	21.10000
75%	7.000000	7.000000	21.600000	26.40000
max	9.000000	9.000000	40.200000	46.70000

Categorical variables:

```
data.describe(include=['O'])
```

	Date	Location	WindGustDir	WindDir9am	WindDir3pm
RainToday	\				

count	145460	145460	135134	134894	141232
142199					
unique	3436	49	16	16	16
2					
top	2015-10-19	Canberra	W	N	SE
No					
freq	49	3436	9915	11758	10838
110319					

	RainTomorrow
count	142193
unique	2
top	No
freq	110316

## Some plots

```
import matplotlib.pyplot as pyplot
```

```
%matplotlib inline
```

### Temperature

```
fig, axs = pyplot.subplots(2,2)
```

```
axs[0,0].set_title('MinTemp histogram')
axs[0,0].hist(data['MinTemp'])
```

```
axs[0,1].set_title('MaxTemp histogram')
axs[0,1].hist(data['MaxTemp'])
```

```
axs[1,0].set_title('Temp9am histogram')
axs[1,0].hist(data['Temp9am'])
```

```
axs[1,1].set_title('Temp3pm histogram')
axs[1,1].hist(data['Temp3pm'])
```

```
/usr/lib/python3/dist-packages/numpy/lib/histograms.py:824:
```

```
RuntimeWarning: invalid value encountered in greater_equal
```

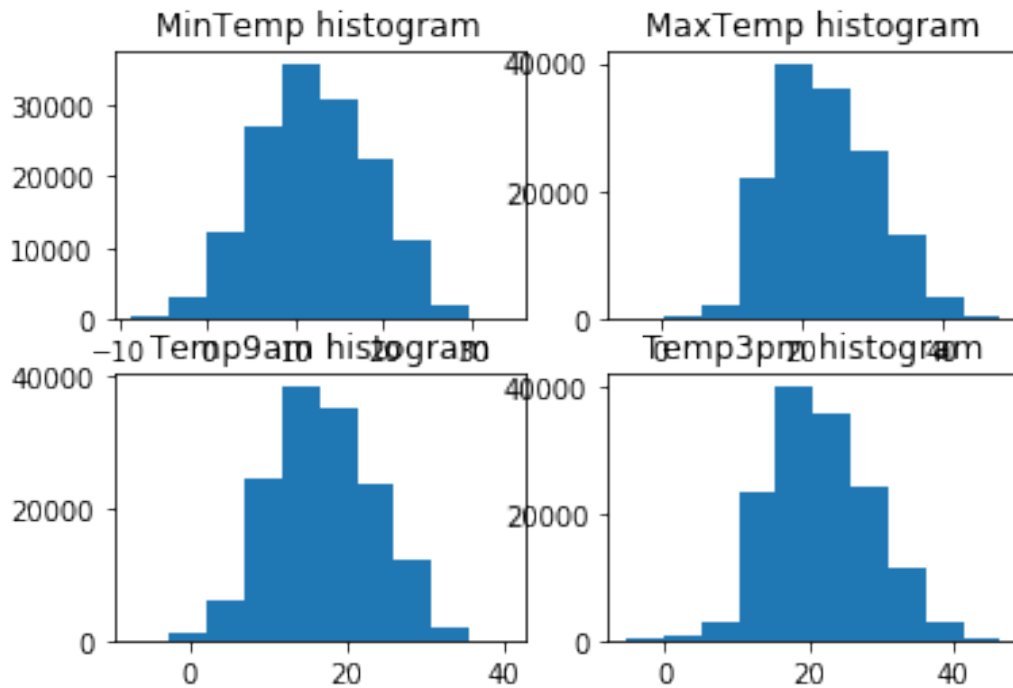
```
    keep = (tmp_a >= first_edge)
```

```
/usr/lib/python3/dist-packages/numpy/lib/histograms.py:825:
```

```
RuntimeWarning: invalid value encountered in less_equal
```

```
    keep &= (tmp_a <= last_edge)
```

```
(array([ 172.,  634., 3112., 23341., 40168., 35761., 24036.,
11314.,
        3030.,  283.]),
 array([-5.4 , -0.19,  5.02, 10.23, 15.44, 20.65, 25.86, 31.07, 36.28,
        41.49, 46.7 ]),
 <a list of 10 Patch objects>)
```



### Rainfall, Evaporation

The data for these features are very skewed. Can apply a log transformation on Evaporation to make it less skewed. It may be more suitable to use RainToday in place of Rainfall.

```
import numpy as np
```

```
def log_transformation(data):
    return data.apply(np.log1p)
```

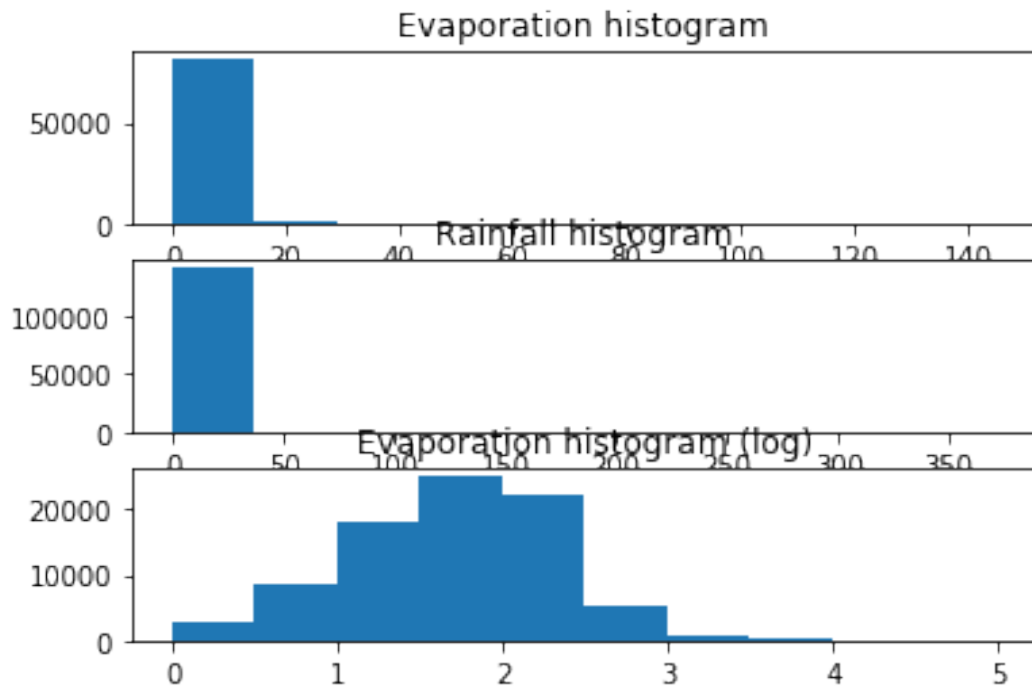
```
fig, axs = pyplot.subplots(3,1)
```

```
axs[0].set_title('Evaporation histogram')
axs[0].hist(data['Evaporation'])
```

```
axs[1].set_title('Rainfall histogram')
axs[1].hist(data['Rainfall'])
```

```
axs[2].set_title('Evaporation histogram (log)')
axs[2].hist(log_transformation(data['Evaporation']))
```

```
(array([2.6450e+03, 8.4680e+03, 1.8182e+04, 2.5154e+04, 2.2130e+04,
        5.3470e+03, 5.5600e+02, 1.4200e+02, 4.5000e+01, 1.0000e+00]),
 array([0.         , 0.49836066, 0.99672132, 1.49508199, 1.99344265,
        2.49180331, 2.99016397, 3.48852464, 3.9868853 , 4.48524596,
        4.98360662])),
 <a list of 10 Patch objects>)
```



### Sunshine

```
fig, axs = pyplot.subplots(1,1)
```

```
axs.set_title('Sunshine')
```

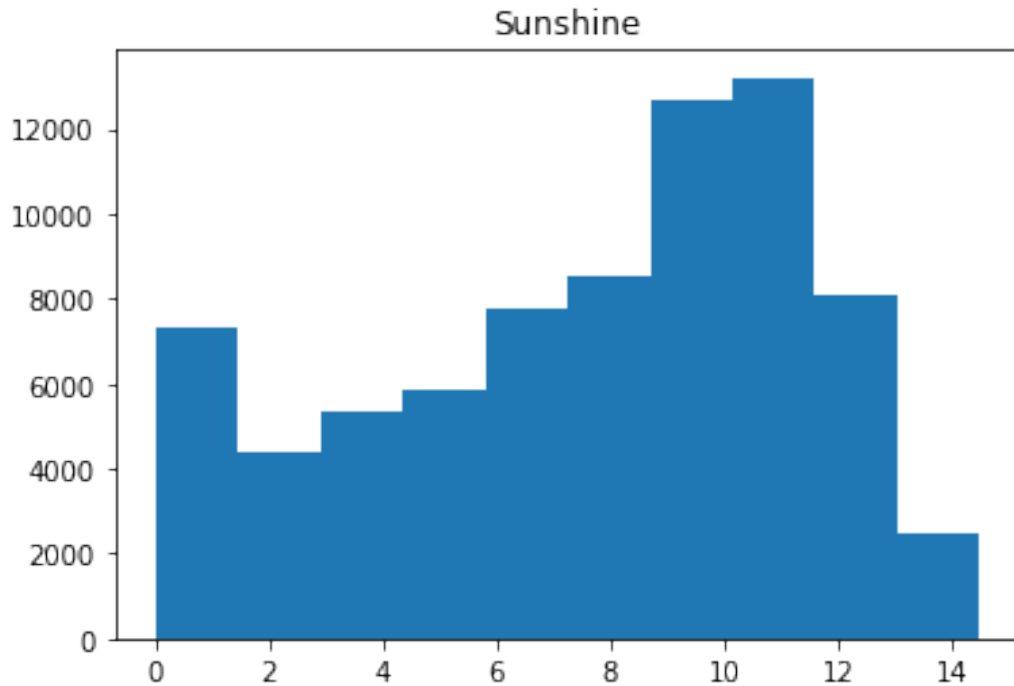
```
axs.hist(data['Sunshine'])
```

```
(array([ 7297.,  4373.,  5314.,  5879.,  7737.,  8548., 12685.,
        13206.,
```

```
        8093.,  2493.] ),
```

```
array([ 0.   ,  1.45,  2.9  ,  4.35,  5.8  ,  7.25,  8.7  , 10.15, 11.6  ,
        13.05, 14.5 ] ),
```

```
<a list of 10 Patch objects>)
```



## Wind

```
# WindGustSpeed WindSpeed9am WindSpeed3pm
```

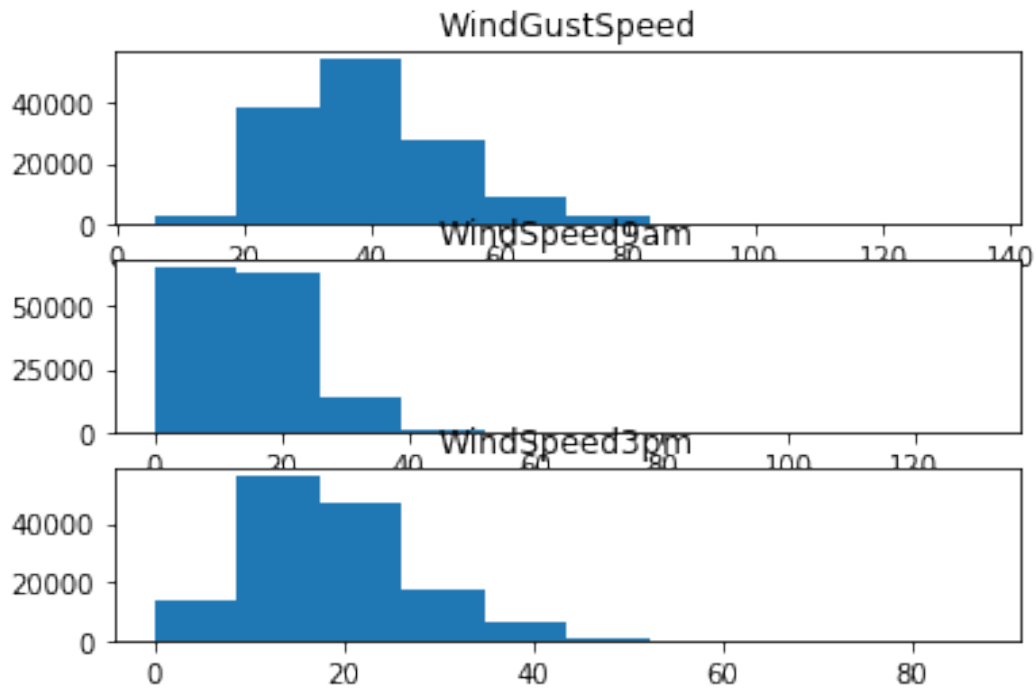
```
fig, axs = pyplot.subplots(3,1)
```

```
axs[0].set_title('WindGustSpeed')
axs[0].hist(data['WindGustSpeed'])
```

```
axs[1].set_title('WindSpeed9am')
axs[1].hist(data['WindSpeed9am'])
```

```
axs[2].set_title('WindSpeed3pm')
axs[2].hist(data['WindSpeed3pm'])
```

```
(array([1.4103e+04, 5.6370e+04, 4.7134e+04, 1.7637e+04, 5.8450e+03,
        1.0870e+03, 1.6000e+02, 5.3000e+01, 6.0000e+00, 3.0000e+00]),
 array([ 0. ,  8.7, 17.4, 26.1, 34.8, 43.5, 52.2, 60.9, 69.6, 78.3,
        87. ]),
 <a list of 10 Patch objects>)
```



## Humidity

```
# Humidity9am Humidity3pm
```

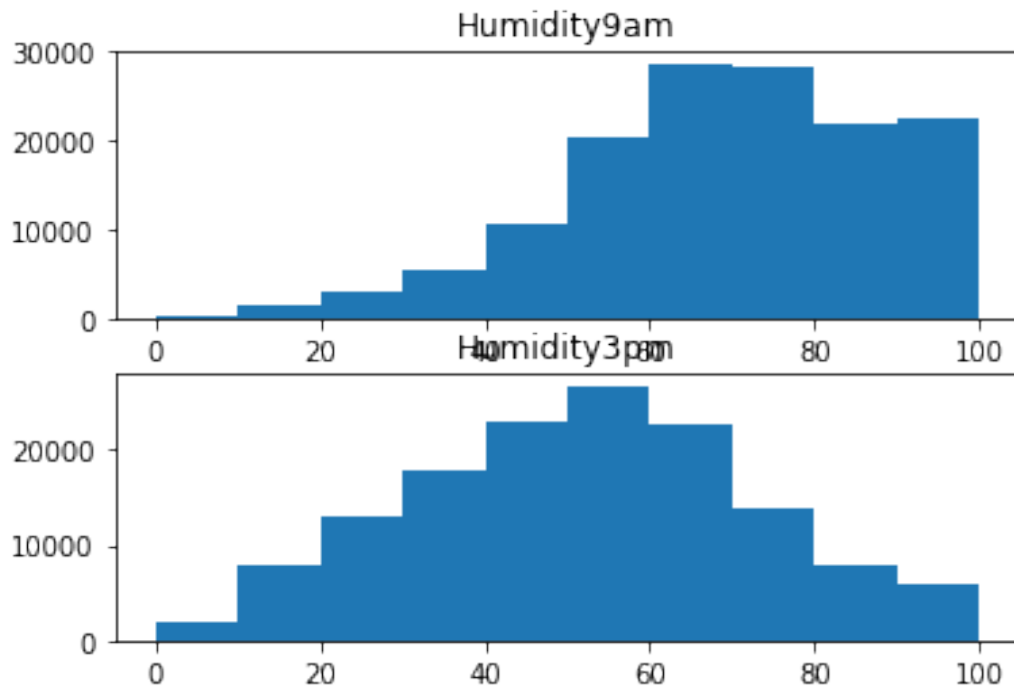
```
fig, axs = pyplot.subplots(2,1)
```

```
axs[0].set_title('Humidity9am')
axs[0].hist(data['Humidity9am'])
```

```
axs[1].set_title('Humidity3pm')
axs[1].hist(data['Humidity3pm'])
```

```
(array([ 1846.,  8040., 13030., 17975., 23028., 26748., 22752.,
        13931.,      7800.,  5803.]),
 array([ 0., 10., 20., 30., 40., 50., 60., 70., 80., 90.,
        100.]),
 <a list of 10 Patch objects>)
```





### Pressure

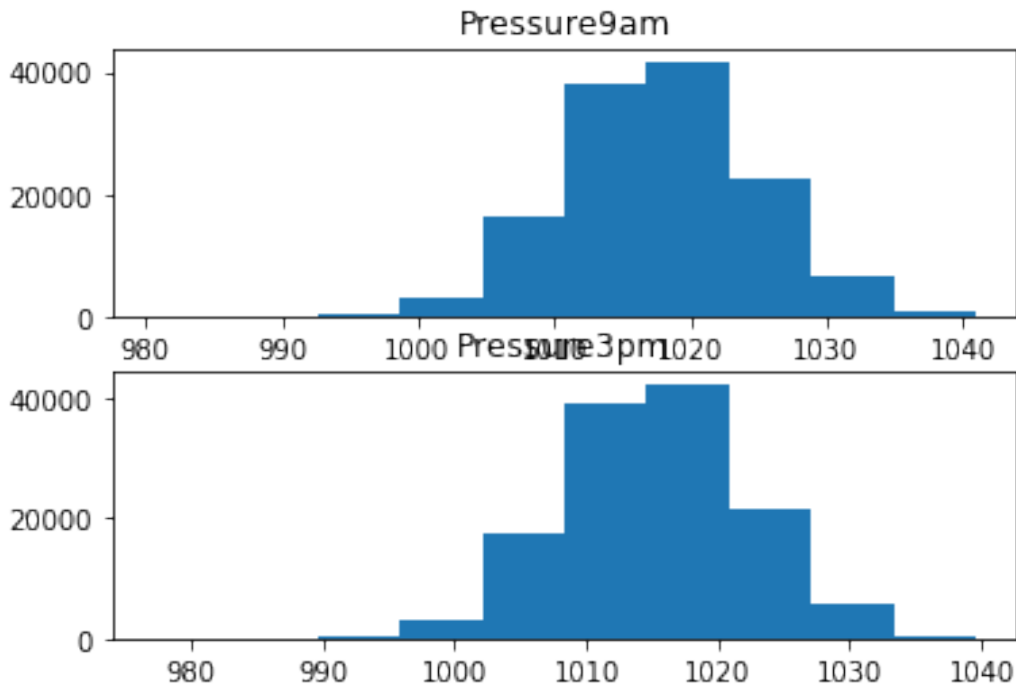
```
# Pressure9am    Pressure3pm
```

```
fig, axs = pyplot.subplots(2,1)
```

```
axs[0].set_title('Pressure9am')
axs[0].hist(data['Pressure9am'])
```

```
axs[1].set_title('Pressure3pm')
axs[1].hist(data['Pressure3pm'])
```

```
(array([1.3000e+01, 8.4000e+01, 5.1200e+02, 2.9200e+03, 1.7509e+04,
        3.9194e+04, 4.2330e+04, 2.1521e+04, 5.7970e+03, 5.5200e+02]),
 array([ 977.1 ,  983.35,  989.6 ,  995.85, 1002.1 , 1008.35, 1014.6 ,
        1020.85, 1027.1 , 1033.35, 1039.6 ]),
 <a list of 10 Patch objects>)
```



## Clouds

Cloud9am and Cloud3pm appear to be bimodal.

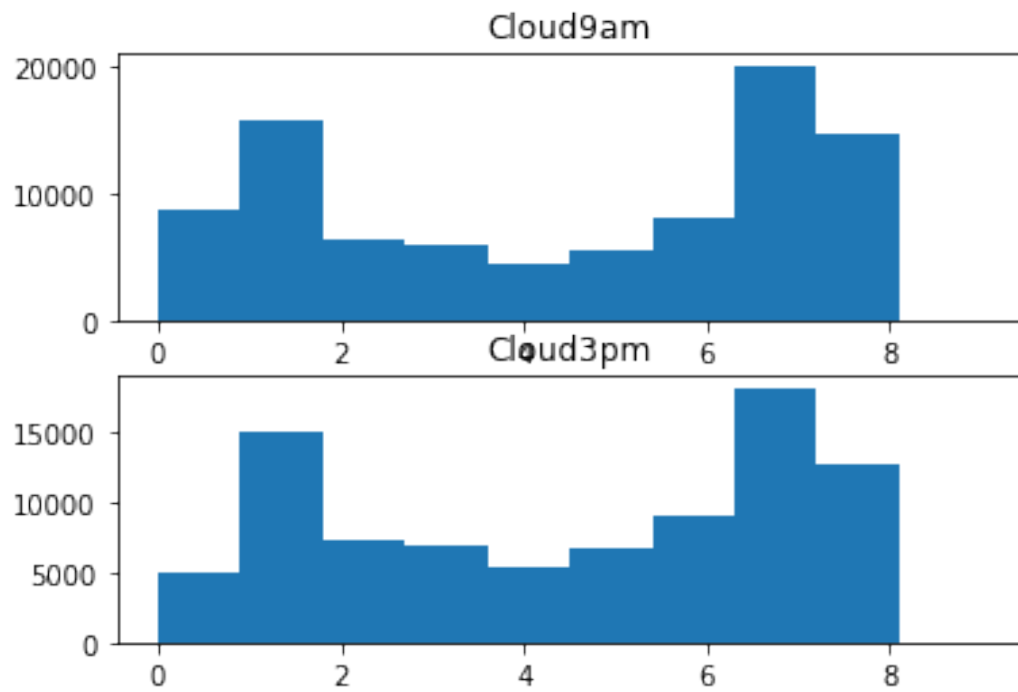
```
# Cloud9am Cloud3pm
```

```
fig, axs = pyplot.subplots(2,1)
```

```
axs[0].set_title('Cloud9am')
axs[0].hist(data['Cloud9am'])
```

```
axs[1].set_title('Cloud3pm')
axs[1].hist(data['Cloud3pm'])
```

```
(array([4.9740e+03, 1.4976e+04, 7.2260e+03, 6.9210e+03, 5.3220e+03,
        6.8150e+03, 8.9780e+03, 1.8229e+04, 1.2660e+04, 1.0000e+00]),
 array([0. , 0.9, 1.8, 2.7, 3.6, 4.5, 5.4, 6.3, 7.2, 8.1, 9. ]),
 <a list of 10 Patch objects>)
```



## Statistical analysis

Done in R Markdown file.