cargar el conjunto de datos de entrenamiento

	In []:	<pre>import pandas as pd</pre>												
		<pre>bike_data = pd.read_csv('daily-bike-share.csv') bike_data.head()</pre>												
	Out[]:		instant	dteday	season	yr	mnth	holiday	weekday	workingday	weathersit	temp	aten	
		0	1	1/1/2011	1	0	1	0	6	0	2	0.344167	0.3636	
		1	2	1/2/2011	1	0	1	0	0	0	2	0.363478	0.3537	
		2	3	1/3/2011	1	0	1	0	1	1	1	0.196364	0.1894	
		3	4	1/4/2011	1	0	1	0	2	1	1	0.200000	0.2121	
		4	5	1/5/2011	1	0	1	0	3	1	1	0.226957	0.2292	
4													•	
<pre>In []: bike_data['day'] = pd.DatetimeIndex(bike_data['dteday']).day bike_data.head(32)</pre>														

Out[]:		instant	dteday	season	yr	mnth	holiday	weekday	workingday	weathersit	temp	at
	0	1	1/1/2011	1	0	1	0	6	0	2	0.344167	0.36
	1	2	1/2/2011	1	0	1	0	0	0	2	0.363478	0.35
	2	3	1/3/2011	1	0	1	0	1	1	1	0.196364	0.18
	3	4	1/4/2011	1	0	1	0	2	1	1	0.200000	0.21
	4	5	1/5/2011	1	0	1	0	3	1	1	0.226957	0.22
	5	6	1/6/2011	1	0	1	0	4	1	1	0.204348	0.23
	6	7	1/7/2011	1	0	1	0	5	1	2	0.196522	0.20
	7	8	1/8/2011	1	0	1	0	6	0	2	0.165000	0.16
	8	9	1/9/2011	1	0	1	0	0	0	1	0.138333	0.11
	9	10	1/10/2011	1	0	1	0	1	1	1	0.150833	0.15
	10	11	1/11/2011	1	0	1	0	2	1	2	0.169091	0.19
	11	12	1/12/2011	1	0	1	0	3	1	1	0.172727	0.16
	12	13	1/13/2011	1	0	1	0	4	1	1	0.165000	0.15
	13	14	1/14/2011	1	0	1	0	5	1	1	0.160870	0.18
	14	15	1/15/2011	1	0	1	0	6	0	2	0.233333	0.24
	15	16	1/16/2011	1	0	1	0	0	0	1	0.231667	0.23
	16	17	1/17/2011	1	0	1	1	1	0	2	0.175833	0.17
	17	18	1/18/2011	1	0	1	0	2	1	2	0.216667	0.23
	18	19	1/19/2011	1	0	1	0	3	1	2	0.292174	0.29
	19	20	1/20/2011	1	0	1	0	4	1	2	0.261667	0.25
	20	21	1/21/2011	1	0	1	0	5	1	1	0.177500	0.15
	21	22	1/22/2011	1	0	1	0	6	0	1	0.059130	0.07
	22	23	1/23/2011	1	0	1	0	0	0	1	0.096522	0.09
	23	24	1/24/2011	1	0	1	0	1	1	1	0.097391	0.11
	24	25	1/25/2011	1	0	1	0	2	1	2	0.223478	0.23
	25	26	1/26/2011	1	0	1	0	3	1	3	0.217500	0.20
	26	27	1/27/2011	1	0	1	0	4	1	1	0.195000	0.21
	27	28	1/28/2011	1	0	1	0	5	1	2	0.203478	0.22
	28	29	1/29/2011	1	0	1	0	6	0	1	0.196522	0.21
	29	30	1/30/2011	1	0	1	0	0	0	1	0.216522	0.25
	30	31	1/31/2011	1	0	1	0	1	1	2	0.180833	0.18
	31	32	2/1/2011	1	0	2	0	2	1	2	0.192174	0.23
1												•

Out[]:

```
In [ ]: numeric_features = ['temp', 'atemp', 'hum', 'windspeed']
   bike_data[numeric_features + ['rentals']].describe()
```

		temp	atemp	hum	windspeed	rentals
	count	731.000000	731.000000	731.000000	731.000000	731.000000
	mean	0.495385	0.474354	0.627894	0.190486	848.176471
	std	0.183051	0.162961	0.142429	0.077498	686.622488
	min	0.059130	0.079070	0.000000	0.022392	2.000000
	25%	0.337083	0.337842	0.520000	0.134950	315.500000
	50%	0.498333	0.486733	0.626667	0.180975	713.000000
	75%	0.655417	0.608602	0.730209	0.233214	1096.000000
	max	0.861667	0.840896	0.972500	0.507463	3410.000000

Esto garantiza que los gráficos se muestren en línea en el bloc de notas de Jupyter

Obtener la columna de etiqueta

Crear una figura para 2 subgráficos (2 filas, 1 columna)

Trazar el histograma

Agregar líneas para la media, la mediana y el modo

Trazar el diagrama de caja

Agregar un título a la figura

Mostrar la figura

```
import pandas as pd
import matplotlib.pyplot as plt

// matplotlib inline

// label = bike_data['rentals']

// fig, ax = plt.subplots(2, 1, figsize = (9,12))

// ax[0].hist(label, bins=100)
```

```
ax[0].set_ylabel('Frequency')

#
ax[0].axvline(label.mean(), color='magenta', linestyle='dashed', linewidth=2)
ax[0].axvline(label.median(), color='cyan', linestyle='dashed', linewidth=2)

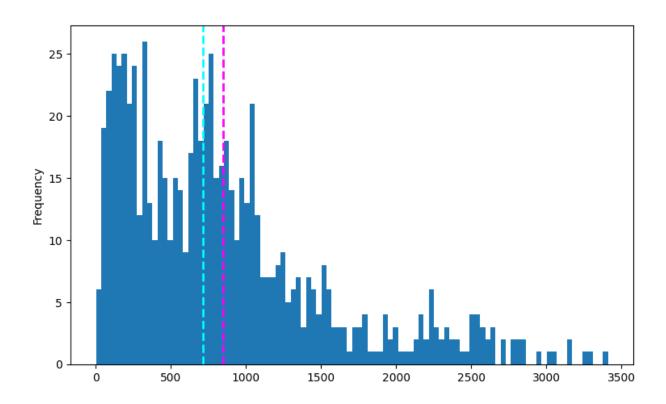
#
ax[1].boxplot(label, vert=False)
ax[1].set_xlabel('Rentals')

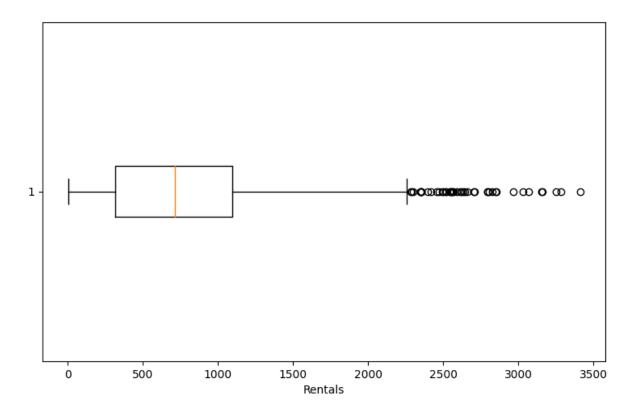
#
fig.suptitle('Rental Distribution')

#
fig.show()
```

C:\Users\lenovo\AppData\Local\Temp\ipykernel_16684\2229708211.py:30: UserWarning: Mat
plotlib is currently using module://matplotlib_inline.backend_inline, which is a nonGUI backend, so cannot show the figure.
 fig.show()

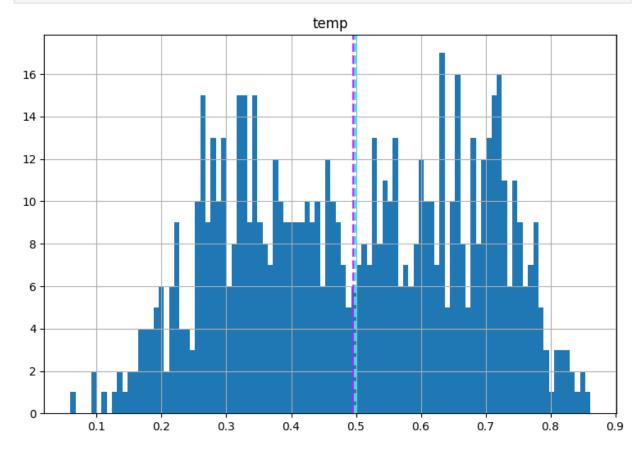
Rental Distribution

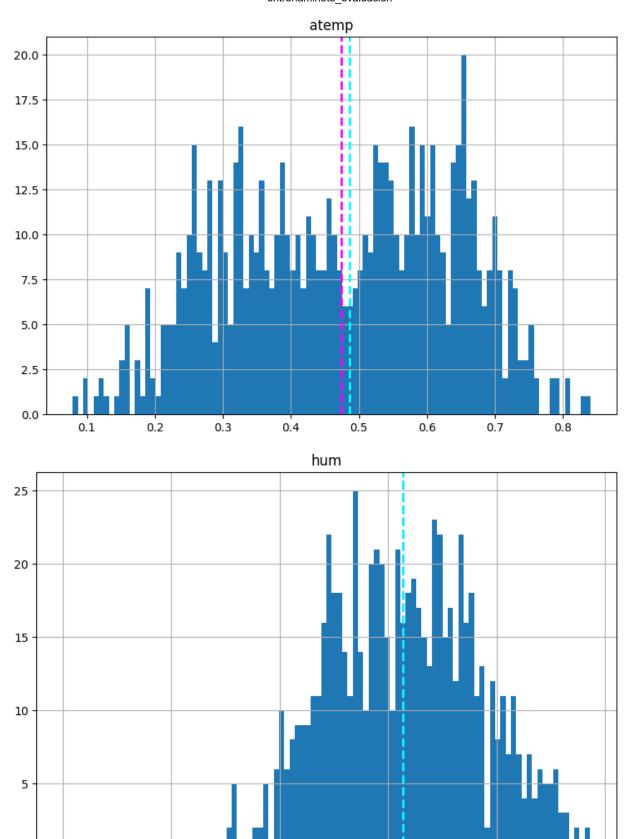




Trazar un histograma para cada entidad numérica

```
In []: #
    for col in numeric_features:
        fig = plt.figure(figsize=(9, 6))
        ax = fig.gca()
        feature = bike_data[col]
        feature.hist(bins=100, ax = ax)
        ax.axvline(feature.mean(), color='magenta', linestyle='dashed', linewidth=2)
        ax.axvline(feature.median(), color='cyan', linestyle='dashed', linewidth=2)
        ax.set_title(col)
    plt.show()
```





0

0.0

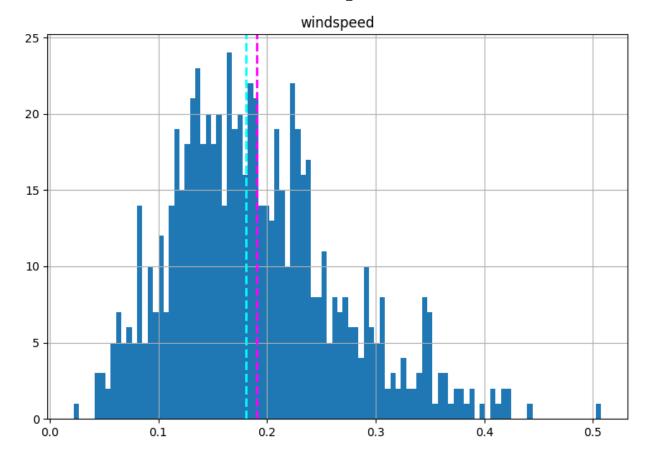
0.2

0.4

0.6

1.0

0.8



Trazar un gráfico de barras para cada recuento de características categóricas

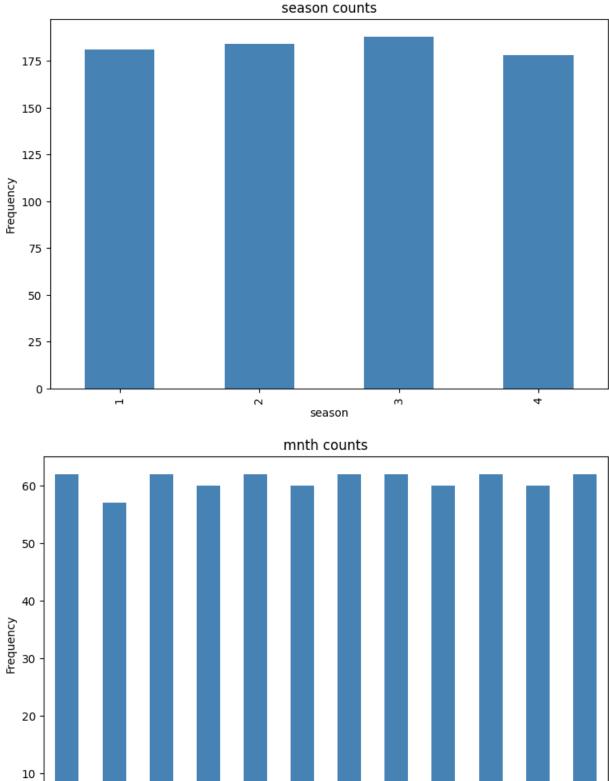
```
import numpy as np

#

categorical_features = ['season','mnth','holiday','weekday','workingday','weathersit',

for col in categorical_features:
    counts = bike_data[col].value_counts().sort_index()
    fig = plt.figure(figsize=(9, 6))
    ax = fig.gca()
    counts.plot.bar(ax = ax, color='steelblue')
    ax.set_title(col + ' counts')
    ax.set_xlabel(col)
    ax.set_ylabel("Frequency")
plt.show()
```



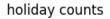


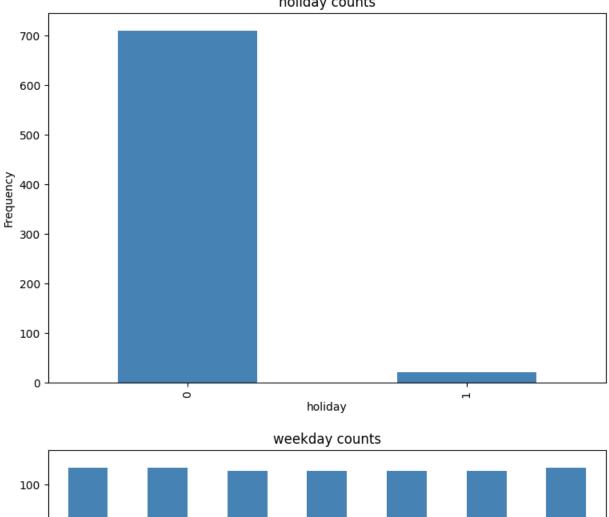
mnth

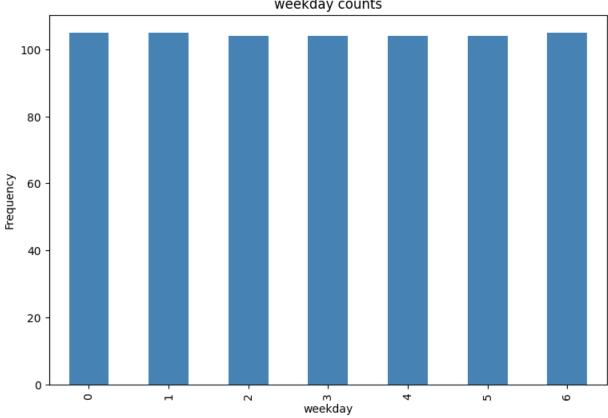
10

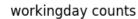
11

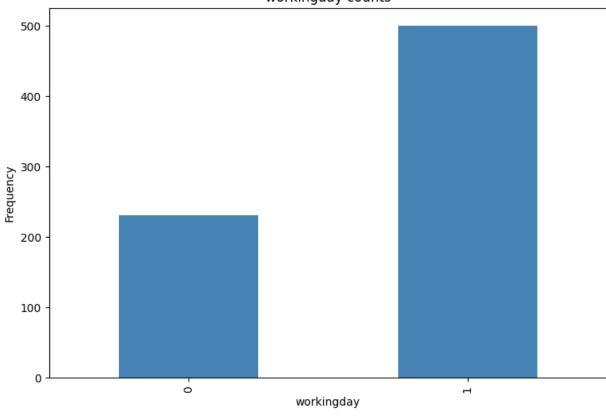
12



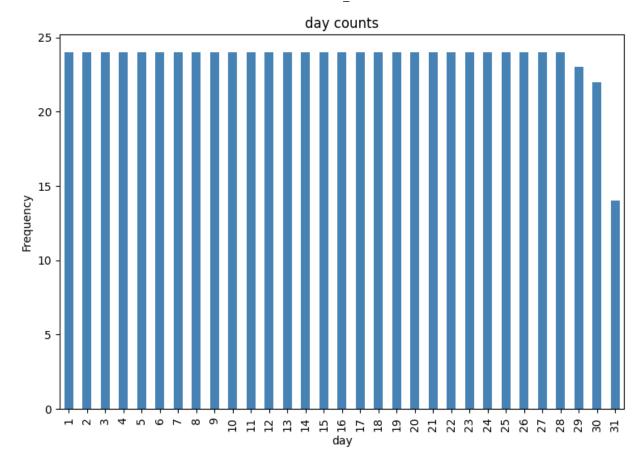






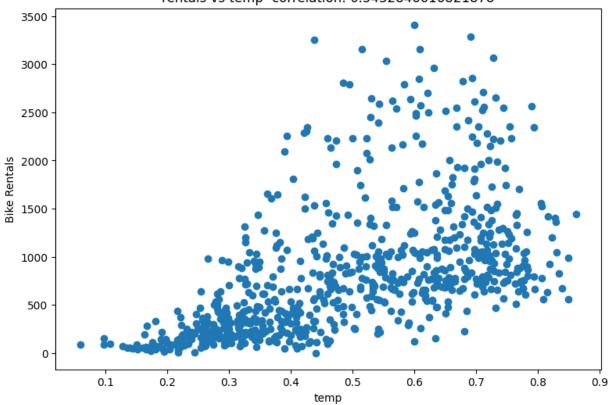


weathersit counts 400 - 300 - 200 - 100 - 200 -

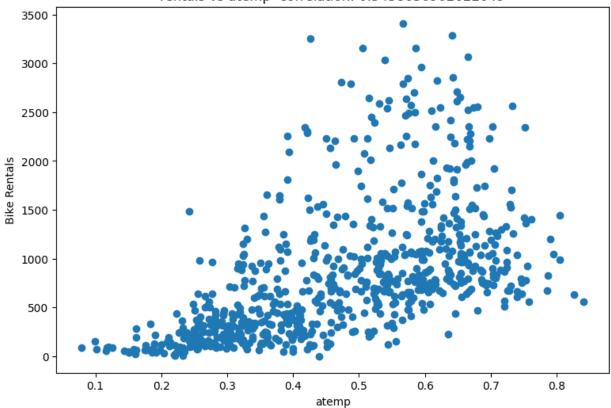


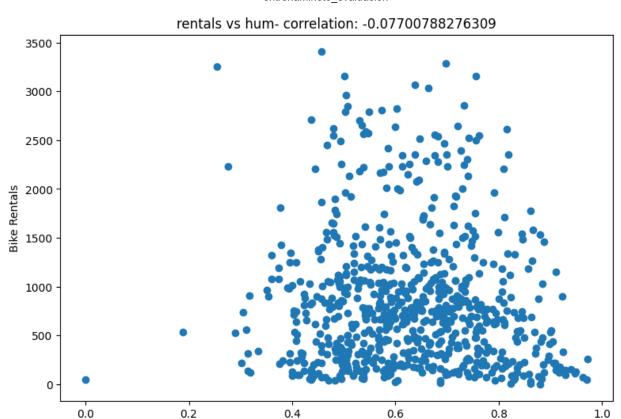
```
In []:
    for col in numeric_features:
        fig = plt.figure(figsize=(9, 6))
        ax = fig.gca()
        feature = bike_data[col]
        label = bike_data['rentals']
        correlation = feature.corr(label)
        plt.scatter(x=feature, y=label)
        plt.xlabel(col)
        plt.ylabel('Bike Rentals')
        ax.set_title('rentals vs ' + col + '- correlation: ' + str(correlation))
    plt.show()
```

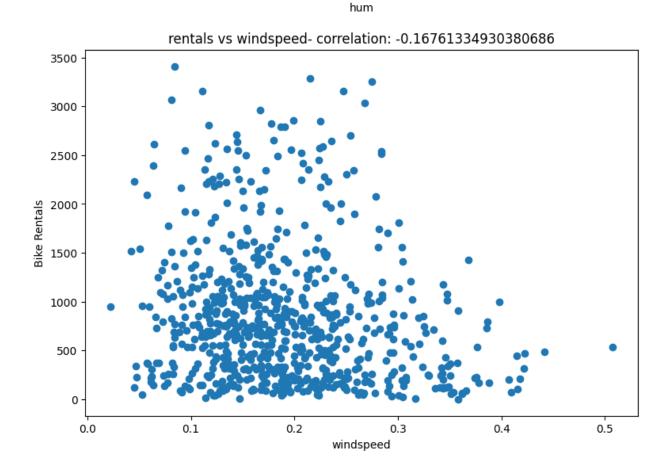
rentals vs temp-correlation: 0.5432846616821878



rentals vs atemp- correlation: 0.5438636902622049



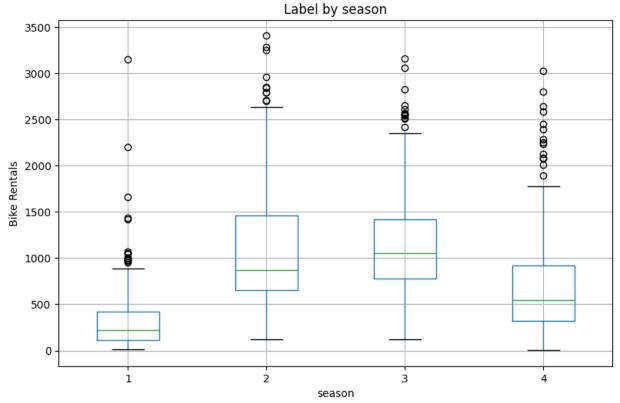




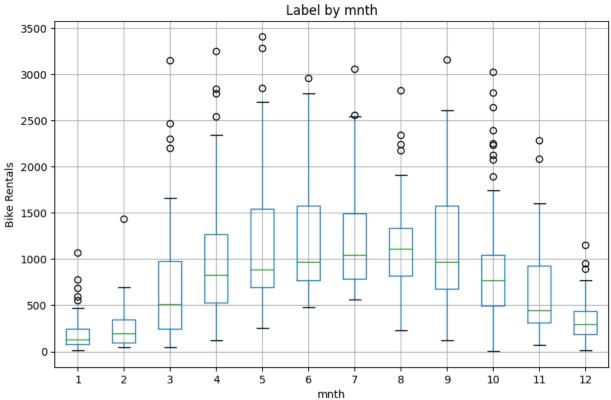
Trazar un diagrama de caja para la etiqueta por cada característica categórica

```
In []: #
    for col in categorical_features:
        fig = plt.figure(figsize=(9, 6))
        ax = fig.gca()
        bike_data.boxplot(column = 'rentals', by = col, ax = ax)
        ax.set_title('Label by ' + col)
        ax.set_ylabel("Bike Rentals")
    plt.show()
```

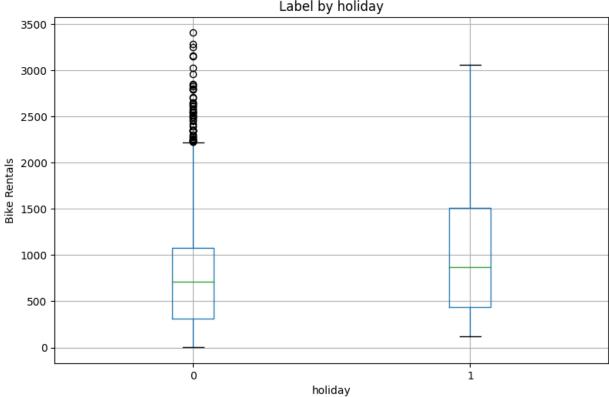
Boxplot grouped by season



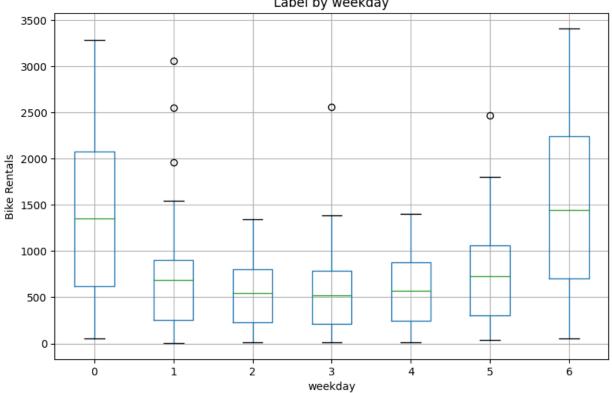
Boxplot grouped by mnth



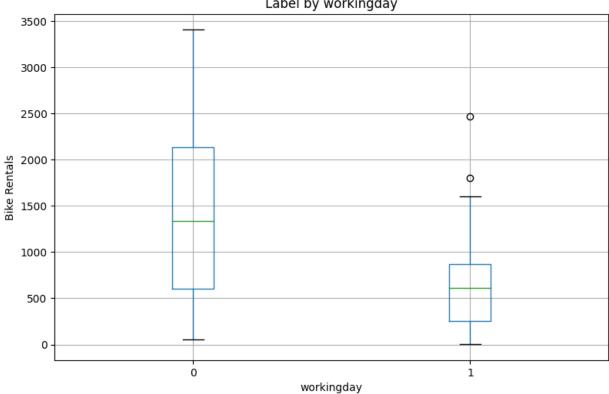
Boxplot grouped by holiday Label by holiday



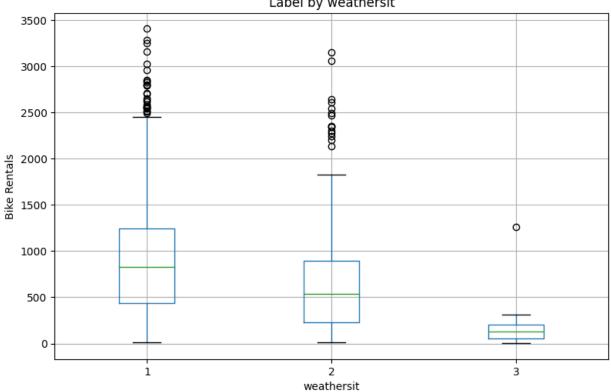
Boxplot grouped by weekday Label by weekday

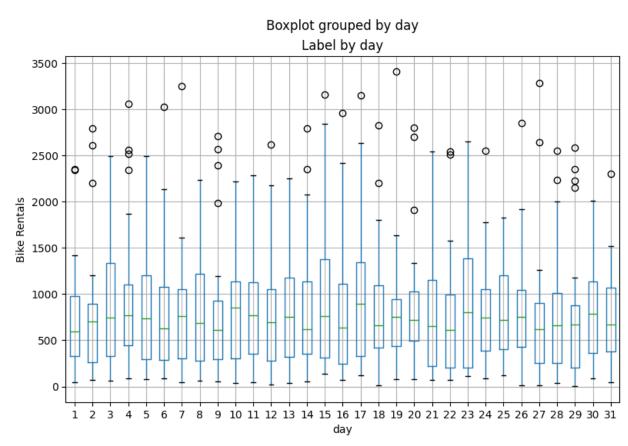


Boxplot grouped by workingday Label by workingday



Boxplot grouped by weathersit Label by weathersit





Características y etiquetas separadas

```
In [ ]: #
X, y = bike_data[['season','mnth', 'holiday','weekday','workingday','weathersit','temp
```

```
print('Features:',X[:10], '\nLabels:', y[:10], sep='\n')
Features:
[[1.
          1.
                                   0.
                                            2.
                                                    0.344167
                  0.
                           6.
 0.363625 0.805833 0.160446 ]
          1. 0.
                                            2.
                                                    0.363478
 0.353739 0.696087 0.248539 ]
                                                    0.196364
 [1.
          1.
                  0.
                           1.
                                   1.
                                            1.
 0.189405 0.437273 0.248309 ]
          1.
                  0.
                           2.
                                            1.
                                                    0.2
 0.212122 0.590435 0.160296 ]
[1.
                  0.
                                            1.
                                                    0.226957
                          3.
 0.22927 0.436957 0.1869
                          1
                                            1.
                                                    0.204348
                  0.
                                   1.
 0.233209 0.518261 0.0895652]
                                            2.
                                                    0.196522
          1.
                  0.
                           5.
 0.208839 0.498696 0.168726 ]
          1.
 [1.
                  0.
                           6.
                                   0.
                                            2.
                                                    0.165
 0.162254 0.535833 0.266804
          1.
                  0.
                          0.
                                   0.
                                            1.
                                                    0.138333
 0.116175 0.434167 0.36195 ]
        1.
                                            1.
                                                    0.150833
                  0.
                           1.
                                   1.
 0.150888 0.482917 0.223267 ]]
Labels:
[331 131 120 108 82 88 148 68 54 41]
```

Divida los datos entre el 70% y el 30% en el conjunto de entrenamiento y el conjunto de prueba

```
In [ ]: from sklearn.model_selection import train_test_split

#
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_state
    print ('Training Set: %d rows\nTest Set: %d rows' % (X_train.shape[0], X_test.shape[0]

    Training Set: 511 rows
    Test Set: 220 rows
```

Entrenar el modelo

Ajustar un modelo de regresión lineal en el conjunto de entrenamiento

```
In []: #
from sklearn.linear_model import LinearRegression

#
model = LinearRegression().fit(X_train, y_train)
print (model)

LinearRegression()

In []: import numpy as np
```

```
predictions = model.predict(X test)
np.set_printoptions(suppress=True)
print('Predicted labels: ', np.round(predictions)[:10])
print('Actual labels : ',y_test[:10])
Predicted labels: [1896. 1184. 1007. -28.
                                            314.
                                                  385.
                                                        475.
                                                              590. 1476.
Actual labels
                  [2418 754 222
                                    47 244
                                            145
                                                  240
                                                       555 3252
                                                                  38]
```

Superponer la línea de regresión

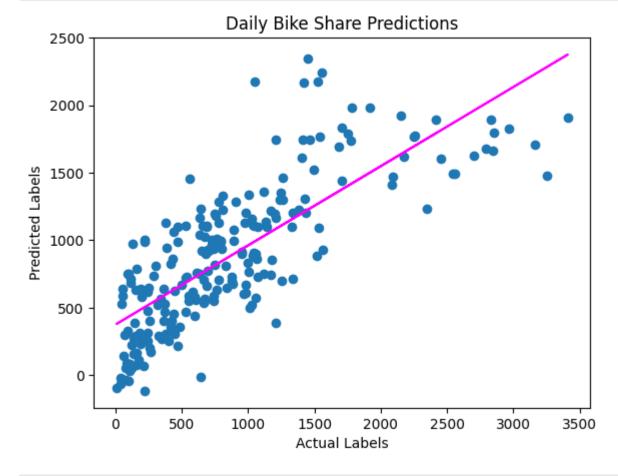
```
In [ ]: import matplotlib.pyplot as plt

%matplotlib inline

plt.scatter(y_test, predictions)
plt.xlabel('Actual Labels')
plt.ylabel('Predicted Labels')
plt.title('Daily Bike Share Predictions')

#

z = np.polyfit(y_test, predictions, 1)
p = np.poly1d(z)
plt.plot(y_test,p(y_test), color='magenta')
plt.show()
```



```
In [ ]: from sklearn.metrics import mean_squared_error, r2_score
    mse = mean_squared_error(y_test, predictions)
    print("MSE:", mse)
```

```
rmse = np.sqrt(mse)
print("RMSE:", rmse)

r2 = r2_score(y_test, predictions)
print("R2:", r2)
```

MSE: 201972.55947035595 RMSE: 449.4135728595165 R2: 0.604045473691919