

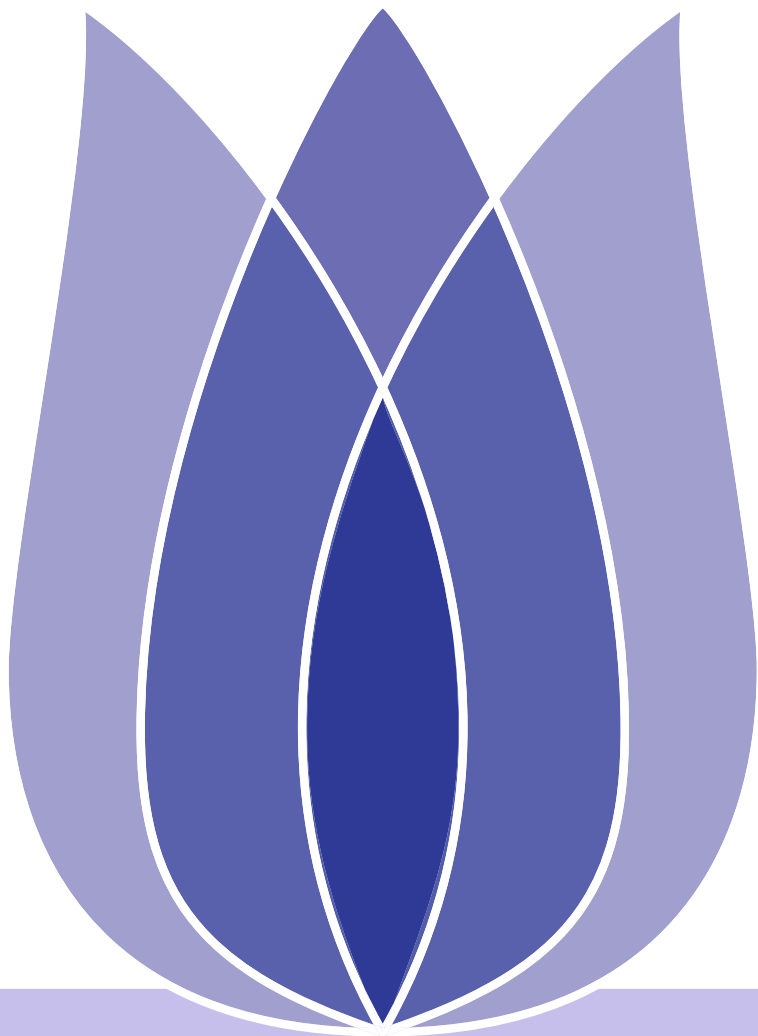


# Bike Sharing Demand

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# Introduction

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Conclusion

- The bike-sharing system is a way to rent bikes through a city-wide network of kiosk locations, automatically gaining membership, renting and returning bikes. People can rent a bike from one place and return it to another as needed.
- Historical car records combine date, weather, temperature, humidity and other factors to predict the bike-sharing program needs in Washington.



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## ■ Descriptive statistics of the data

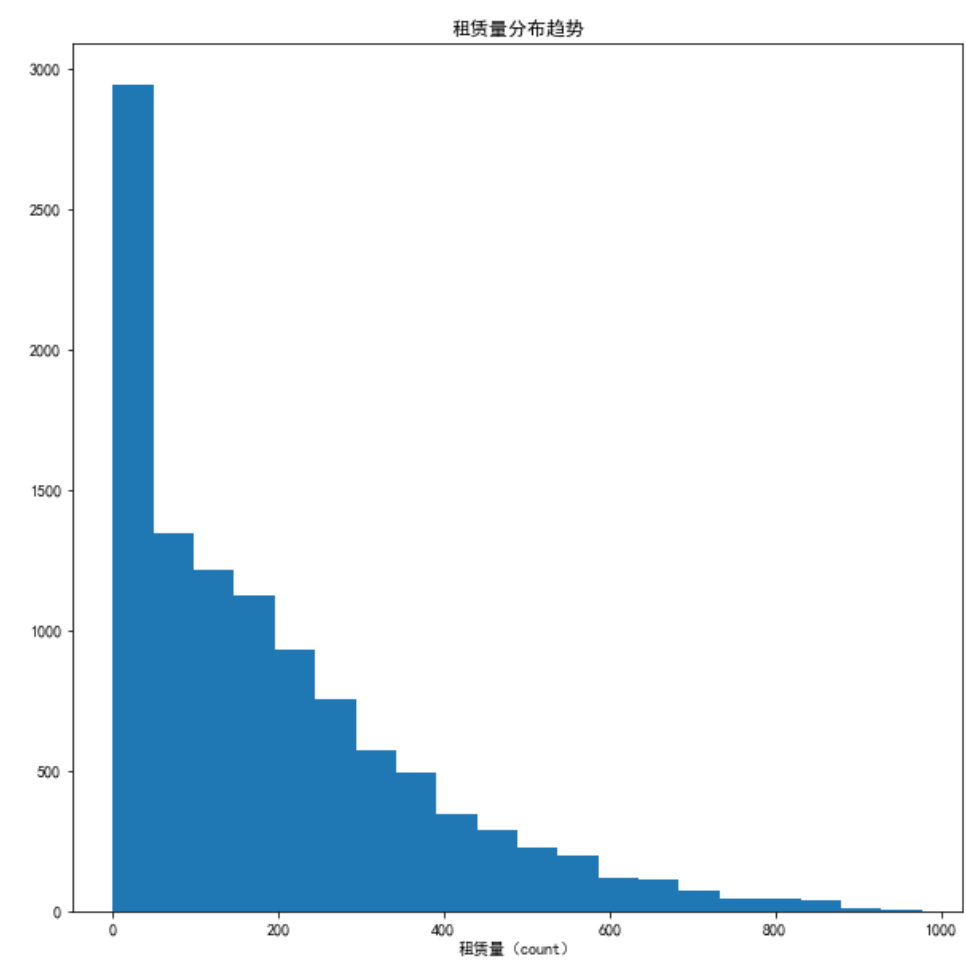
|       | season       | holiday      | workingday   | weather      | temp        | atemp        | humidity     | windspeed    | casual       | registered   | coun         |
|-------|--------------|--------------|--------------|--------------|-------------|--------------|--------------|--------------|--------------|--------------|--------------|
| count | 10886.000000 | 10886.000000 | 10886.000000 | 10886.000000 | 10886.00000 | 10886.000000 | 10886.000000 | 10886.000000 | 10886.000000 | 10886.000000 | 10886.000000 |
| mean  | 2.506614     | 0.028569     | 0.680875     | 1.418427     | 20.23086    | 23.655084    | 61.886460    | 12.799395    | 36.021955    | 155.552177   | 191.574130   |
| std   | 1.116174     | 0.166599     | 0.466159     | 0.633839     | 7.79159     | 8.474601     | 19.245033    | 8.164537     | 49.960477    | 151.039033   | 181.144450   |
| min   | 1.000000     | 0.000000     | 0.000000     | 1.000000     | 0.82000     | 0.760000     | 0.000000     | 0.000000     | 0.000000     | 0.000000     | 1.000000     |
| 25%   | 2.000000     | 0.000000     | 0.000000     | 1.000000     | 13.94000    | 16.665000    | 47.000000    | 7.001500     | 4.000000     | 36.000000    | 42.000000    |
| 50%   | 3.000000     | 0.000000     | 1.000000     | 1.000000     | 20.50000    | 24.240000    | 62.000000    | 12.998000    | 17.000000    | 118.000000   | 145.000000   |
| 75%   | 4.000000     | 0.000000     | 1.000000     | 2.000000     | 26.24000    | 31.060000    | 77.000000    | 16.997900    | 49.000000    | 222.000000   | 284.000000   |
| max   | 4.000000     | 1.000000     | 1.000000     | 4.000000     | 41.00000    | 45.455000    | 100.000000   | 56.996900    | 367.000000   | 886.000000   | 977.000000   |



# Data Visualization

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- The standard deviation of the number of leases you have to predict at the end is very large. So let's look at the distribution by drawing it.

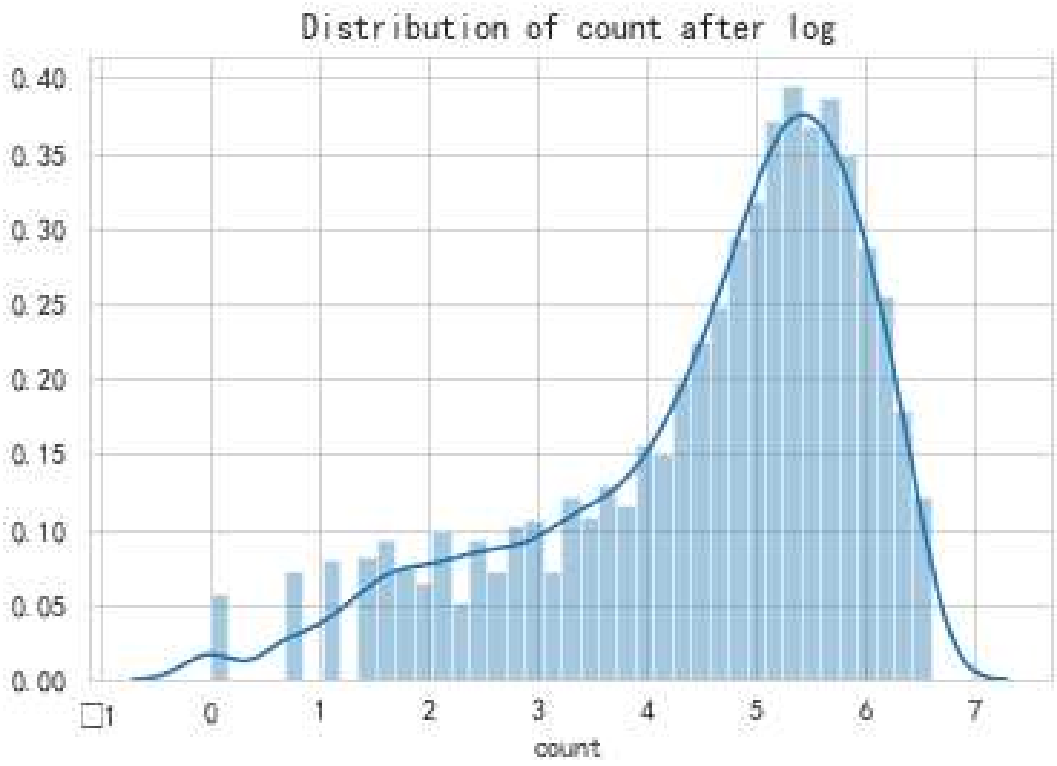
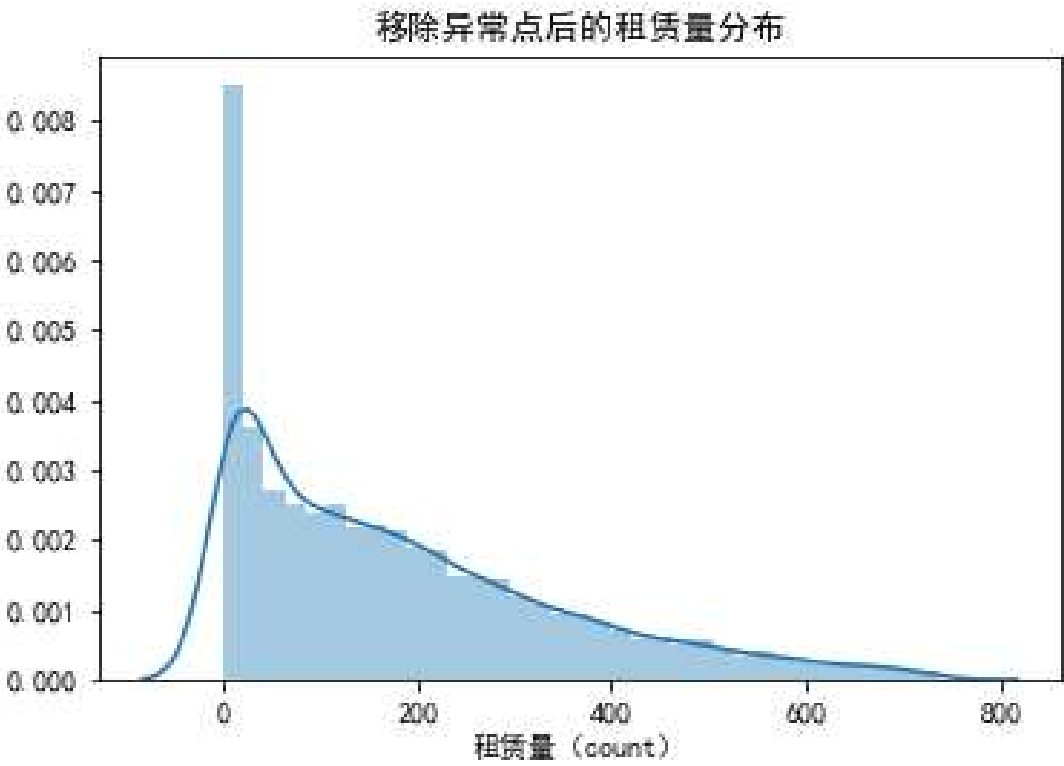




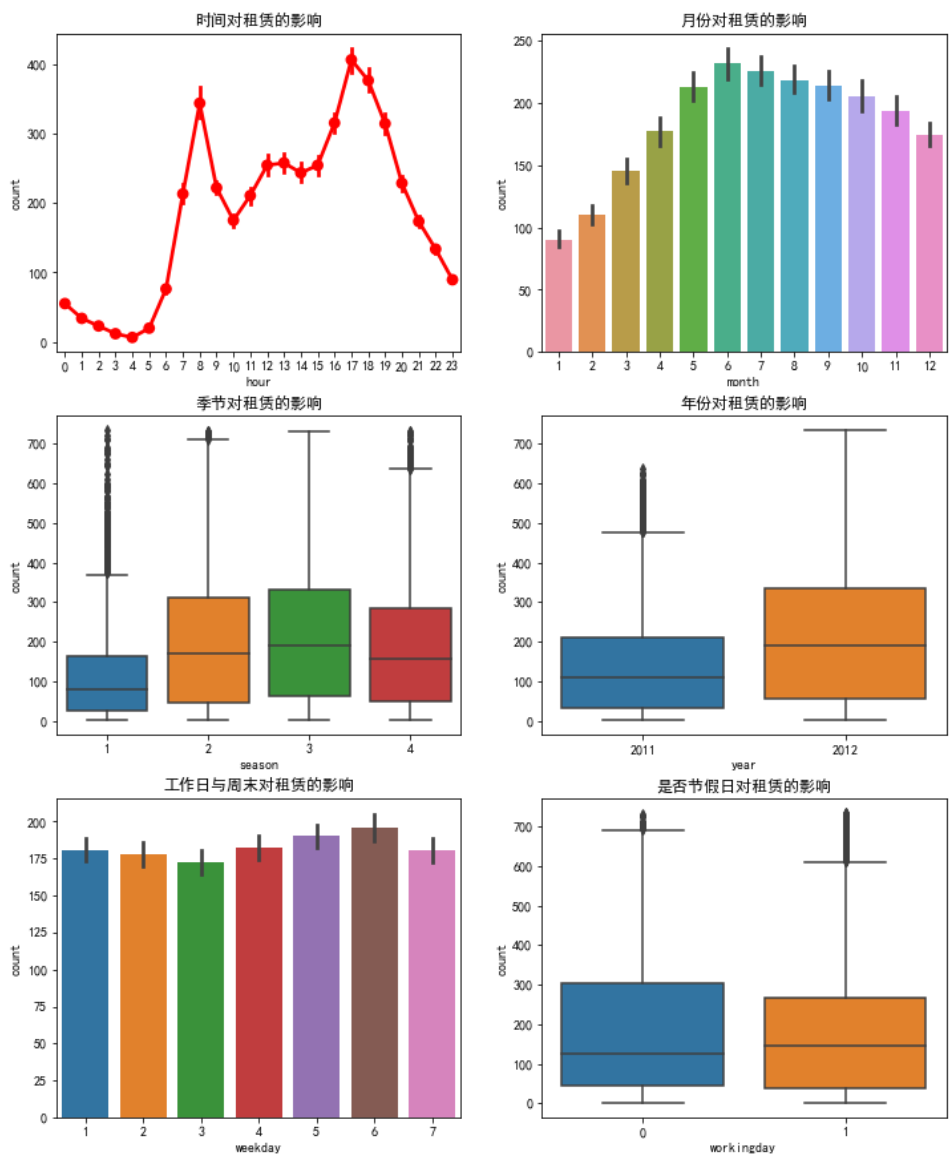
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## ■ Exclude data other than three standards,log of count

|      | season       | holiday      | workingday   | weather      | temp         | atemp        | humidity     | windspeed    | casual       | registered   | cour         |
|------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|--------------|
| ount | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 | 10617.000000 |
| nean | 2.499294     | 0.029104     | 0.676180     | 1.421871     | 20.073588    | 23.490210    | 62.138363    | 12.779423    | 34.301309    | 142.816144   | 177.11745    |
| std  | 1.121325     | 0.168107     | 0.467954     | 0.636097     | 7.779602     | 8.466483     | 19.238023    | 8.175715     | 47.716238    | 128.456579   | 158.26198    |
| min  | 1.000000     | 0.000000     | 0.000000     | 1.000000     | 0.820000     | 0.760000     | 0.000000     | 0.000000     | 0.000000     | 0.000000     | 1.000000     |
| 25%  | 1.000000     | 0.000000     | 0.000000     | 1.000000     | 13.940000    | 16.665000    | 47.000000    | 7.001500     | 4.000000     | 34.000000    | 40.000000    |
| 50%  | 2.000000     | 0.000000     | 1.000000     | 1.000000     | 20.500000    | 24.240000    | 62.000000    | 12.998000    | 16.000000    | 114.000000   | 139.000000   |
| 75%  | 4.000000     | 0.000000     | 1.000000     | 2.000000     | 26.240000    | 31.060000    | 78.000000    | 16.997900    | 46.000000    | 212.000000   | 271.000000   |
| max  | 4.000000     | 1.000000     | 1.000000     | 4.000000     | 41.000000    | 45.455000    | 100.000000   | 56.996900    | 355.000000   | 652.000000   | 663.000000   |



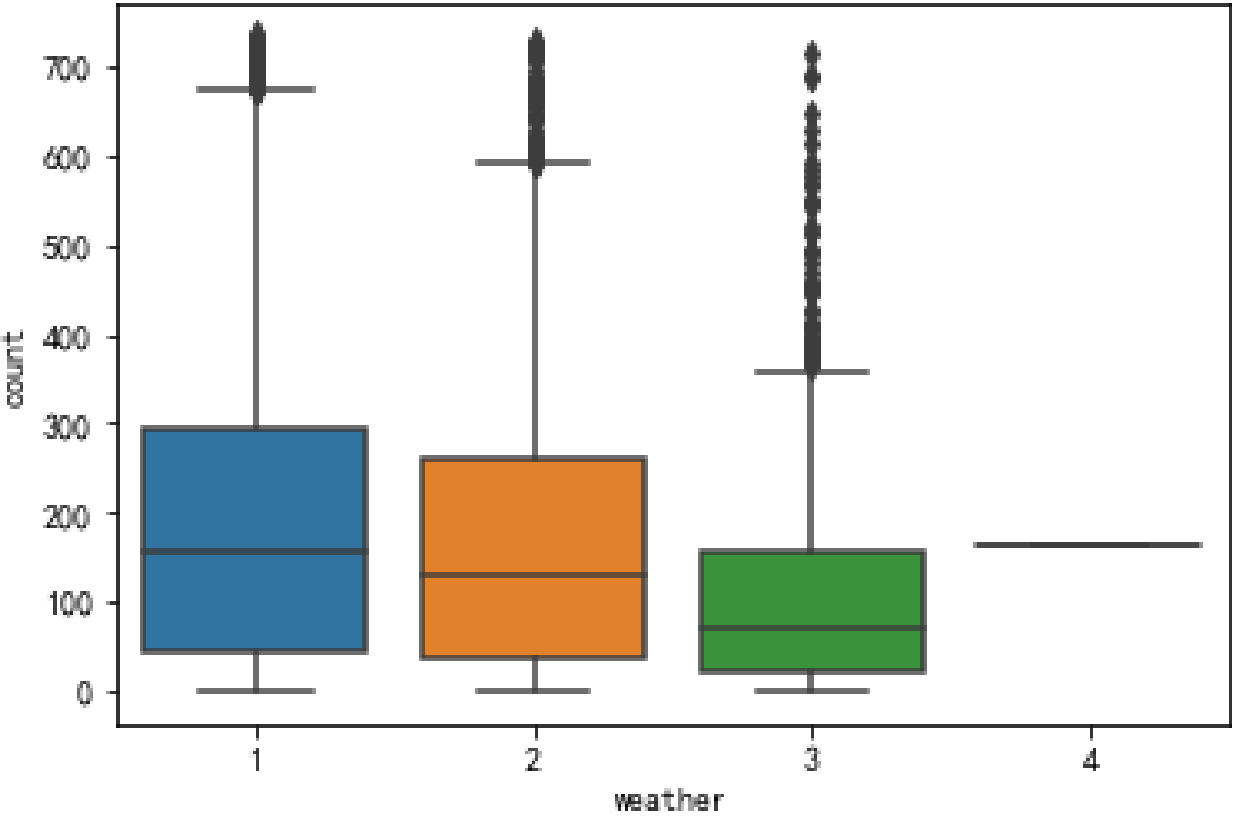
## ■ The impact of hour,month,season,year,weekday,workingday





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## ■ The impact of weather



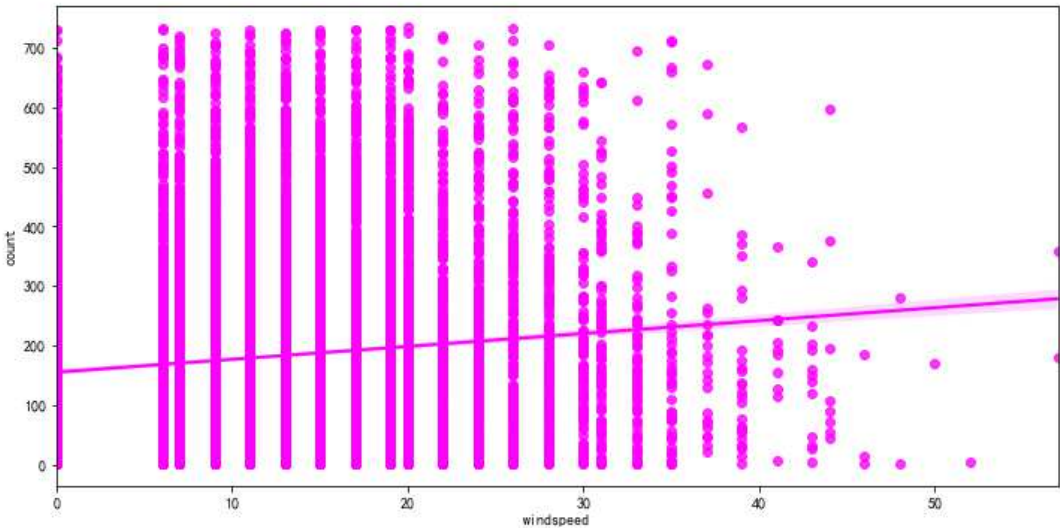
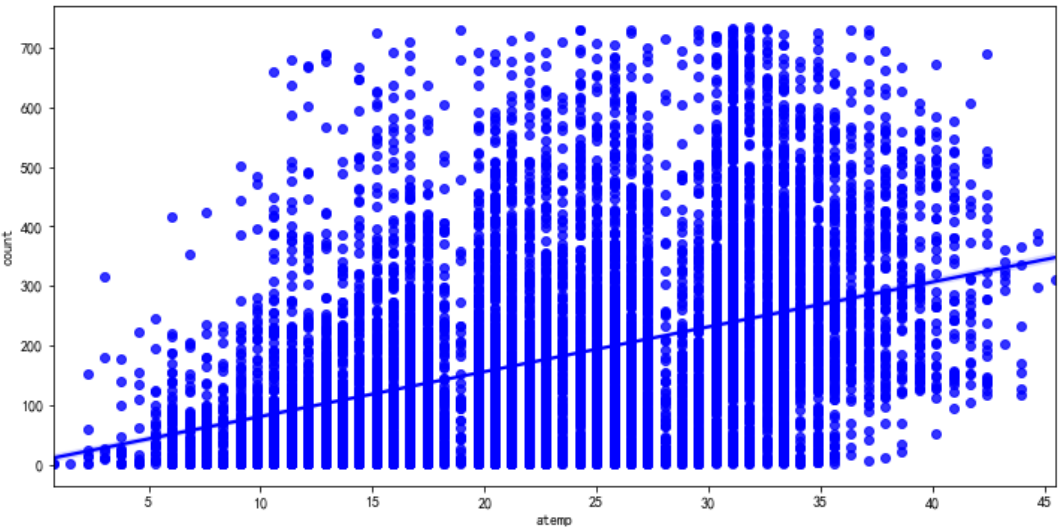
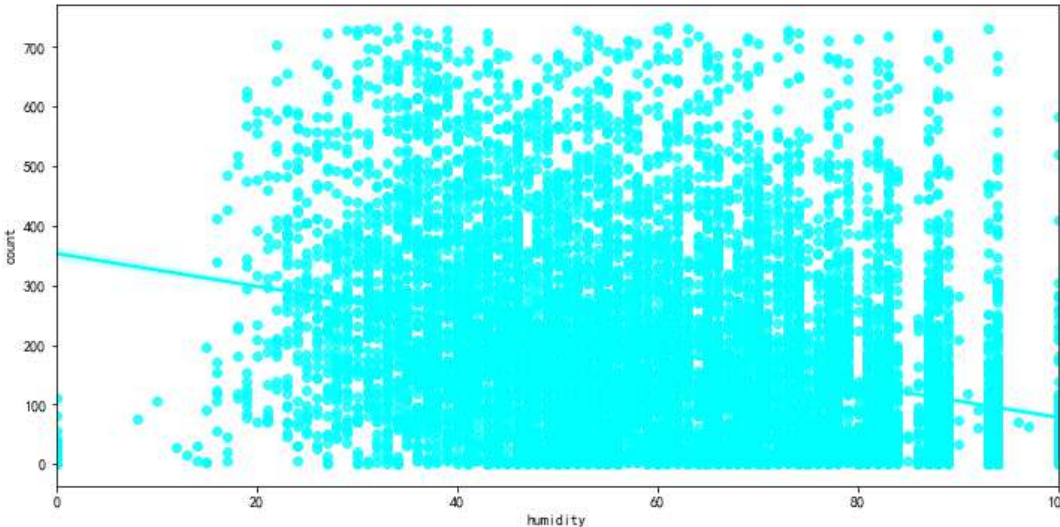
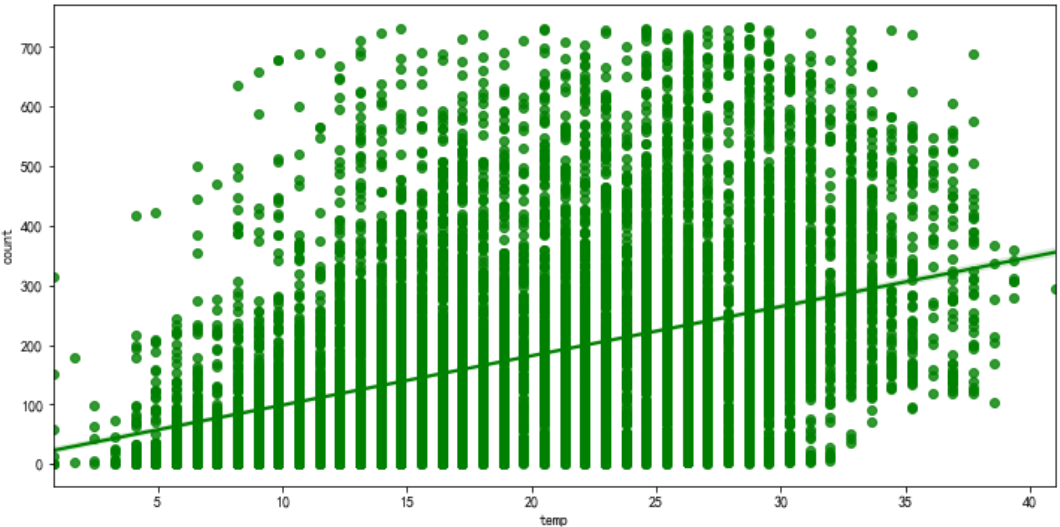




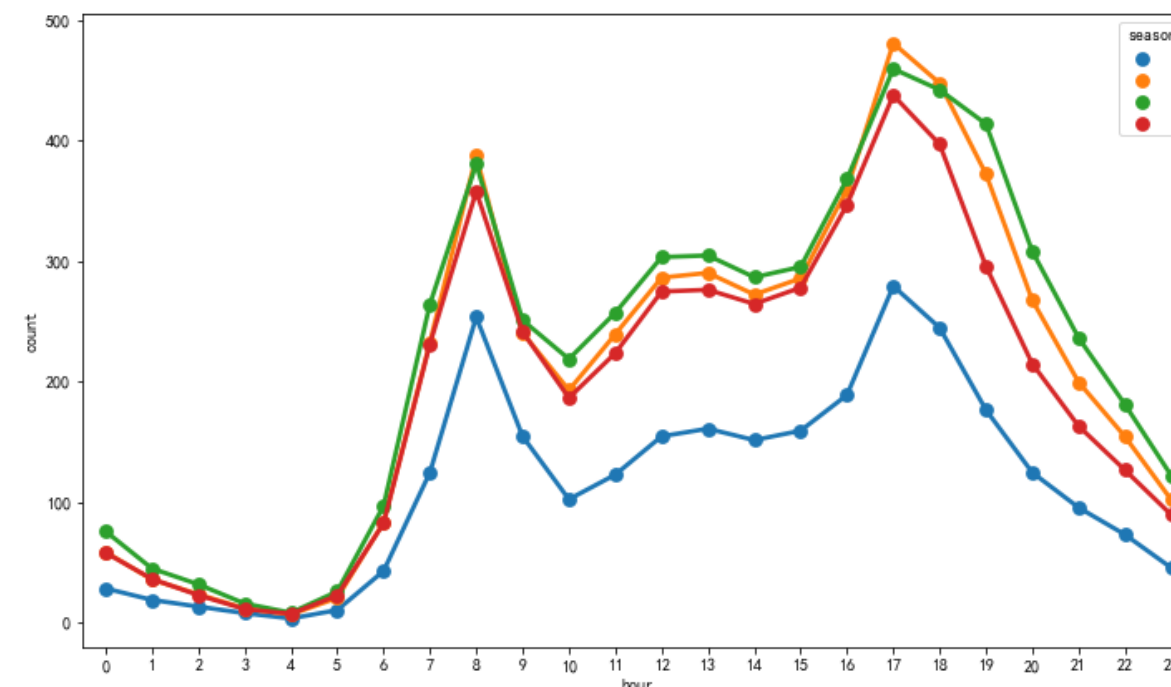
# Data Visualization

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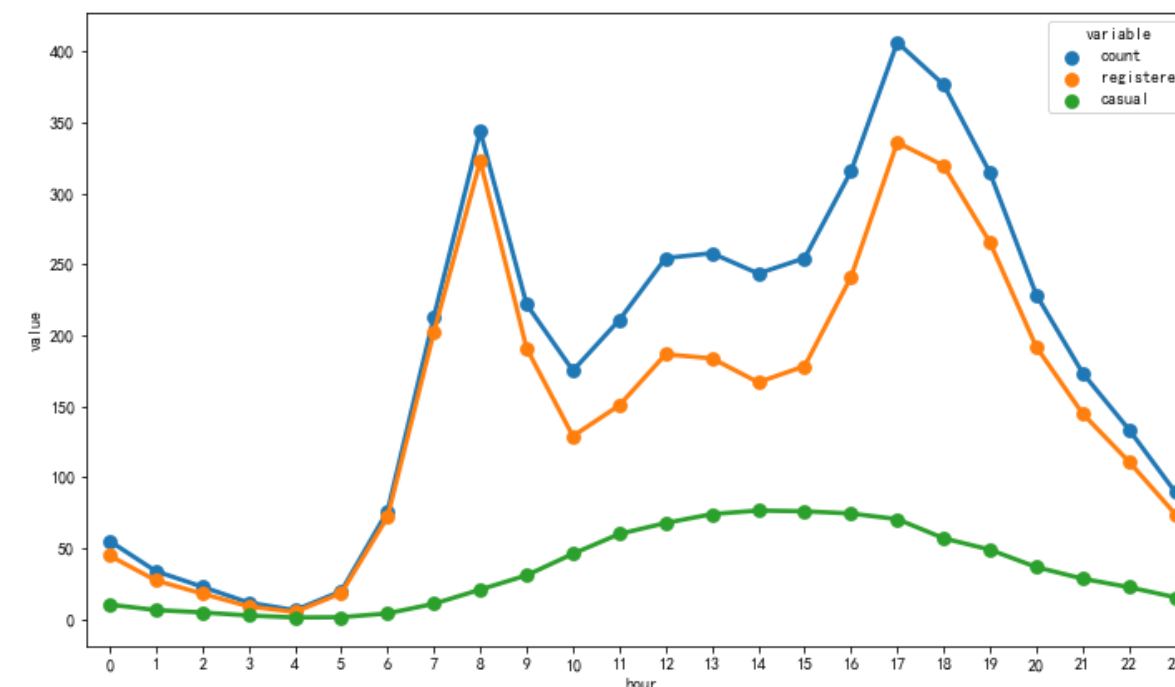
## ■ The impact of temp,atemp,humidity,windspeed



- Impact of season, week, registered and non-registered users on cycling usage trends
- For different times of the day, there is a clear trend in the use of Shared bikes, with two distinct peaks, in line with people's understanding of morning peak and evening peak. The trends were the same for all four seasons, except that usage in spring was slightly lower than in the other three.



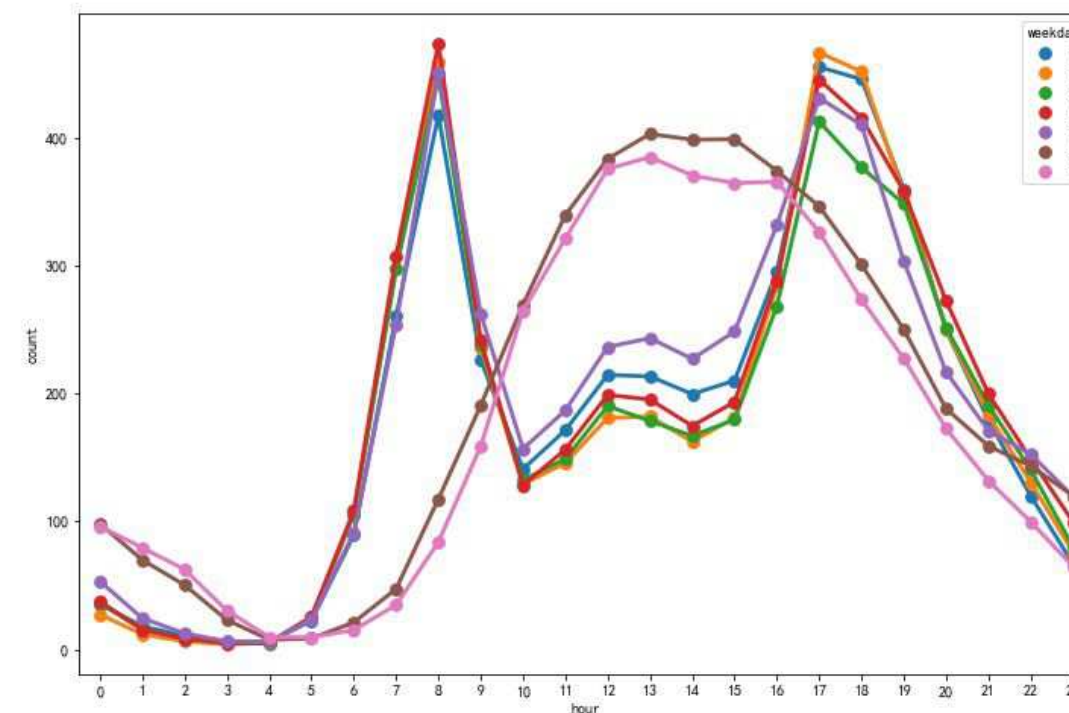
- The usage of registered users accounts for the majority of the total usage, and the trend is consistent with the total usage trend, rather than that of registered users. The usage at different times of the day does not change much, and the trend is similar to the usage trend at weekends.



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- From Monday to Friday, there are two peak usage periods, while on weekends, the usage trend is completely different from that on weekdays. The usage trend changes from bimodal to flat unimodal, and the peak usage period is concentrated at 11-17 o'clock.



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- Draw the thermal diagram of the correlation coefficient



- After all kinds of analysis, we will consider the time periods, temp, humidity, year, month, season, weather grade, windspeed, day of the week, whether there is a workingday or not and whether there is a holiday as characteristic values here.



# Build Model

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- 1. Separate the training set and test set.
- 2. Cross validation is used to determine the optimal parameters.
- 3. View the selected optimal parameters: 'max\_depth': 20, 'n\_estimators': 150
- 4. Apply the optimal parameters to the model, it can be obtained  
Accuracy on test set: 0.6945996275605214



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# Conclution

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- Through this Kaggle project, I practiced by myself to have a deeper understanding of data visualization and to explore the structure and rules of data by means of drawing and tabulating.

