

DAEHYUN KO

OGTIP PROJECT 3

LINEAR REGRESSION ANALYSIS

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PROBLEM STATEMENT

BOOMBIKES DEMAND PREDICTION

Boombikes is a US bike sharing provider who wishes to understand which factors affects the demand of the bikes. And if they can determine future demand predictions.

The main problem statements

- Which variables are significant in predicting the demand for shared bikes.
- How well those variables describe the bike demands

DATA DICTIONARY

- **INSTANT**: RECORD INDEX
- **DTEDAY** : DATE
- **SEASON** : SEASON (1:SPRING, 2:SUMMER, 3:FALL, 4:WINTER)
- **YR** : YEAR (0: 2018, 1:2019)
- **MNTH** : MONTH (1 TO 12)
- **HOLIDAY** : WEATHER DAY IS A HOLIDAY OR NOT
- **WEEKDAY** : DAY OF THE WEEK
- **WORKINGDAY** : IF DAY IS NEITHER WEEKEND NOR HOLIDAY IS 1, OTHERWISE IS 0.
- + **WEATHERSIT** :
 - 1: CLEAR, FEW CLOUDS, PARTLY CLOUDY, PARTLY CLOUDY
 - 2: MIST + CLOUDY, MIST + BROKEN CLOUDS, MIST + FEW CLOUDS, MIST
 - 3: LIGHT SNOW, LIGHT RAIN + THUNDERSTORM + SCATTERED CLOUDS, LIGHT RAIN + SCATTERED CLOUDS
 - 4: HEAVY RAIN + ICE PALLETS + THUNDERSTORM + MIST, SNOW + FOG
- **TEMP** : TEMPERATURE IN CELSIUS
- **ATEMP**: FEELING TEMPERATURE IN CELSIUS
- **HUM**: HUMIDITY
- **WINDSPEED**: WIND SPEED
- **CASUAL**: COUNT OF CASUAL USERS
- **REGISTERED**: COUNT OF REGISTERED USERS
- **CNT**: COUNT OF TOTAL RENTAL BIKES INCLUDING BOTH CASUAL AND REGISTERED

LINEAR REGRESSION ANALYSIS

'TEMP'

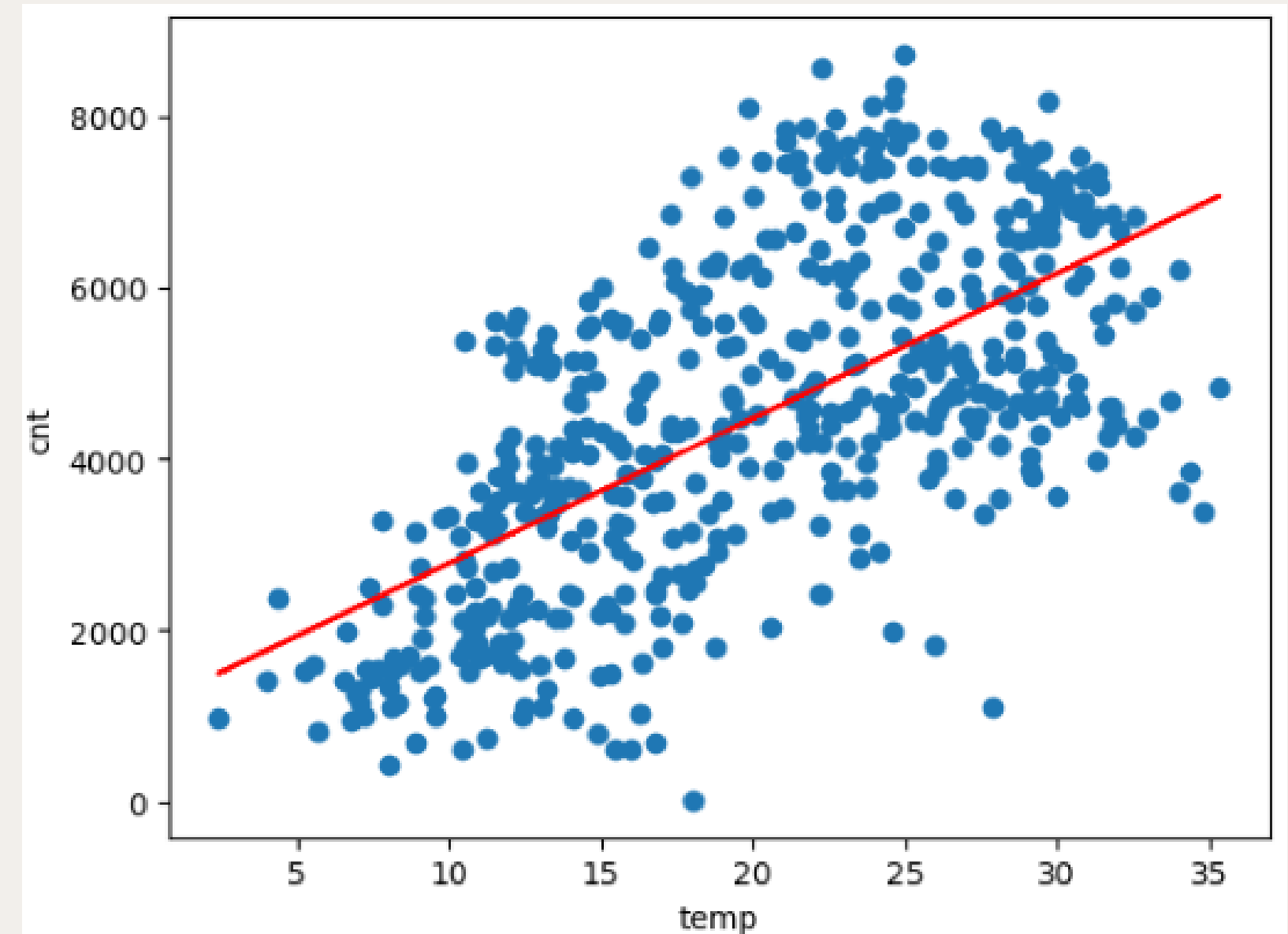
OLS Regression Results

```
=====
Dep. Variable:          cnt    R-squared:                0.414
Model:                  OLS    Adj. R-squared:           0.413
Method:                 Least Squares    F-statistic:        359.1
Date:                  Wed, 10 May 2023    Prob (F-statistic):    5.80e-61
Time:                  00:27:48    Log-Likelihood:       -4450.9
No. Observations:      510    AIC:                8906.
Df Residuals:          508    BIC:                8914.
Df Model:              1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	1088.0439	191.181	5.691	0.000	712.442	1463.646
temp	169.0511	8.921	18.949	0.000	151.524	186.579

```
=====
Omnibus:                6.280    Durbin-Watson:        2.047
Prob(Omnibus):          0.043    Jarque-Bera (JB):      4.555
Skew:                   0.098    Prob(JB):              0.103
Kurtosis:               2.580    Cond. No.              62.0
=====
```

MODEL



- demand increases as outside temperature increases

'WEATHERSIT'

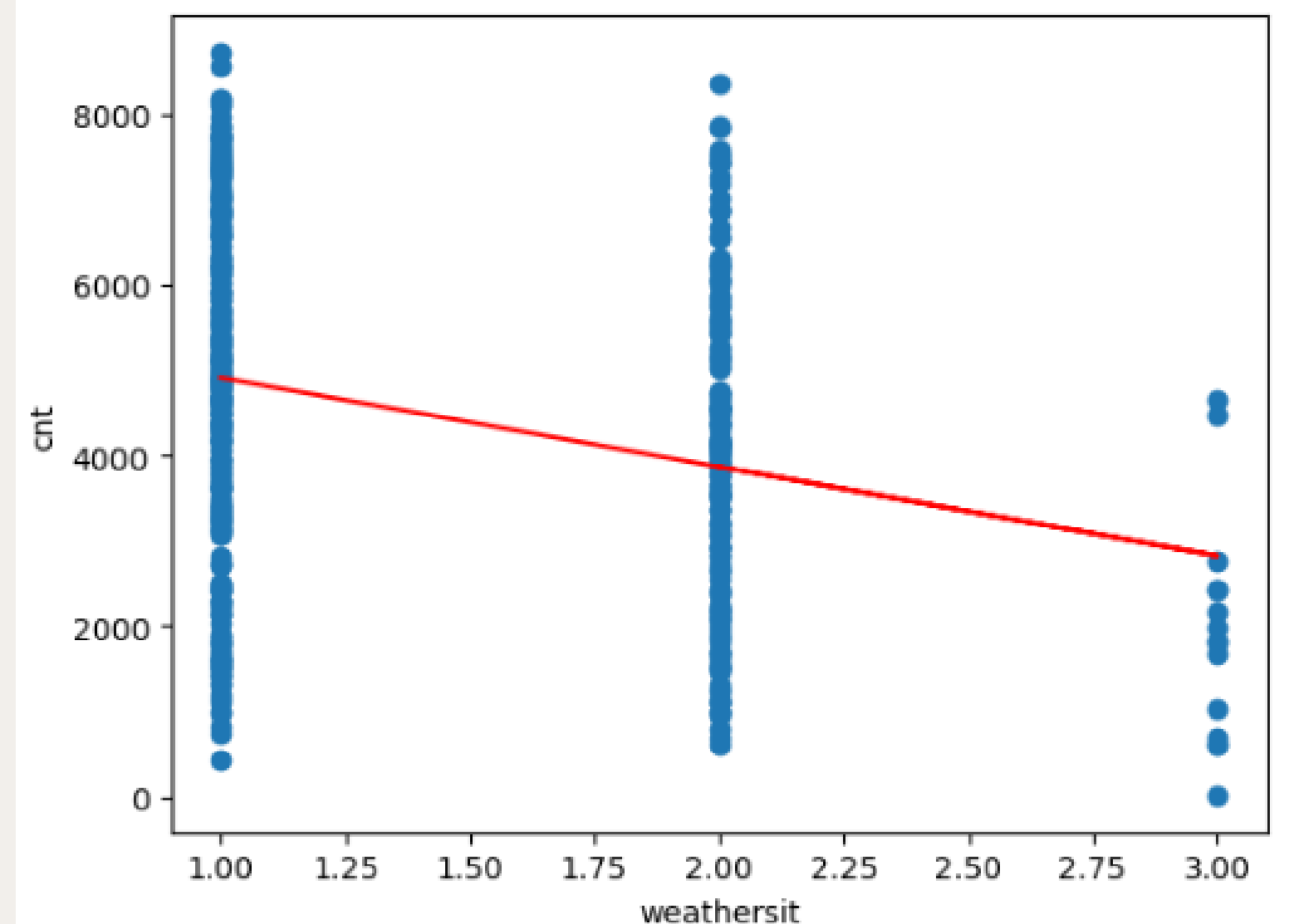
OLS Regression Results

```
=====
Dep. Variable:          cnt    R-squared:          0.085
Model:                  OLS    Adj. R-squared:       0.084
Method:                 Least Squares    F-statistic:      47.47
Date:                  Wed, 10 May 2023    Prob (F-statistic): 1.66e-11
Time:                  00:27:49    Log-Likelihood:    -4564.5
No. Observations:      510    AIC:              9133.
Df Residuals:          508    BIC:              9141.
Df Model:              1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	5947.5767	227.654	26.125	0.000	5500.317	6394.836
weathersit	-1042.2505	151.276	-6.890	0.000	-1339.454	-745.047

```
=====
Omnibus:                27.312    Durbin-Watson:          1.903
Prob(Omnibus):           0.000    Jarque-Bera (JB):       11.363
Skew:                   -0.073    Prob(JB):               0.00341
Kurtosis:                2.283    Cond. No.:               5.80
=====
```

MODEL



- demand is highest on clear days, medium on cloudy days and worst on rainy, snowy days

'HUM'(HUMIDITY)

OLS Regression Results

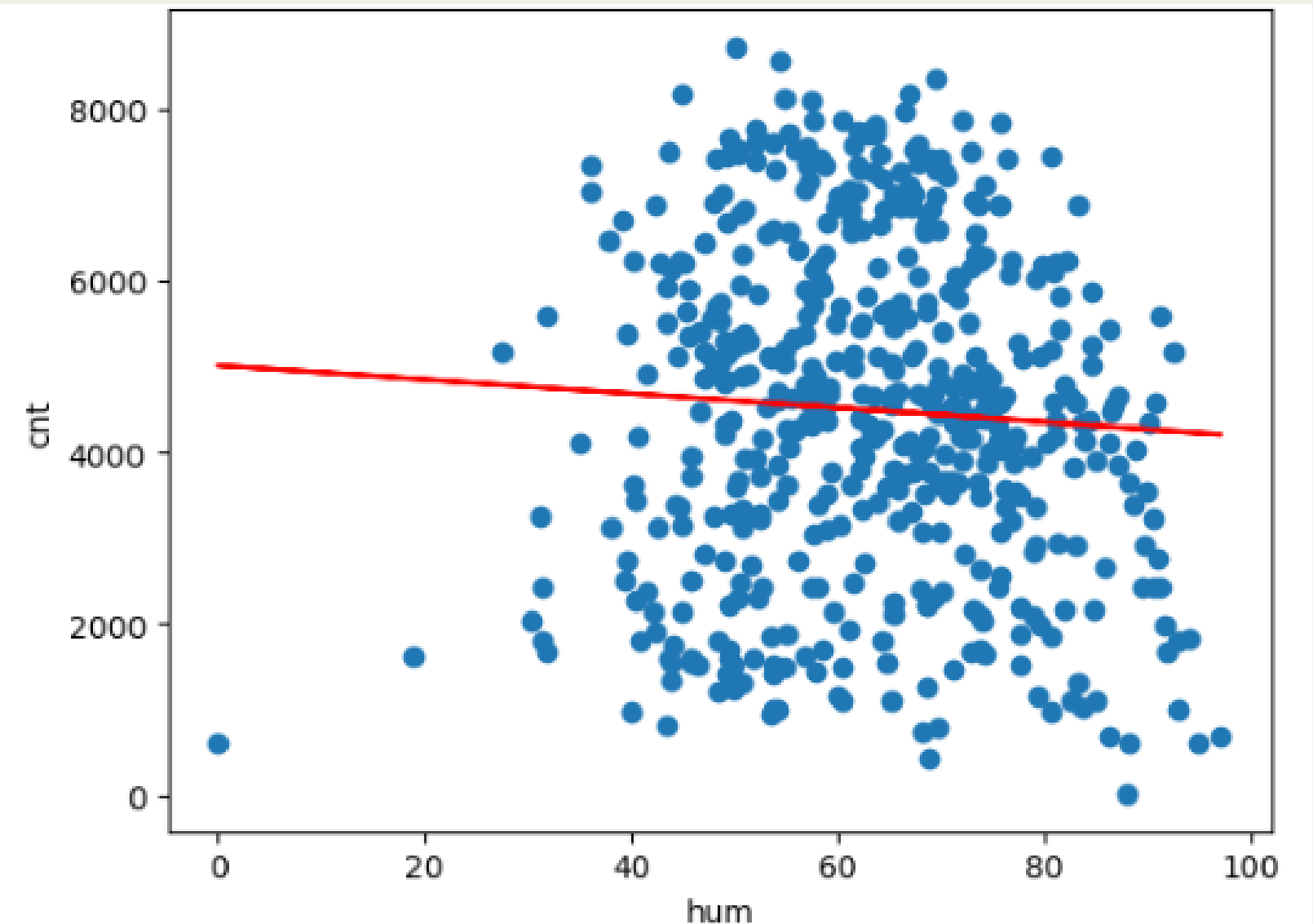
```
=====
Dep. Variable:          cnt    R-squared:                0.004
Model:                  OLS    Adj. R-squared:           0.002
Method:                 Least Squares    F-statistic:        1.835
Date:                   Wed, 10 May 2023    Prob (F-statistic):    0.176
Time:                   00:27:50    Log-Likelihood:       -4586.4
No. Observations:       510    AIC:                 9177.
Df Residuals:           508    BIC:                 9185.
Df Model:                1
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	5008.5098	395.001	12.680	0.000	4232.474	5784.546
hum	-8.2729	6.107	-1.355	0.176	-20.271	3.726

```
=====
Omnibus:                49.348    Durbin-Watson:        1.881
Prob(Omnibus):           0.000    Jarque-Bera (JB):     15.658
Skew:                   -0.076    Prob(JB):              0.000398
Kurtosis:                2.155    Cond. No.              296.
=====
```

Things to note:
p-value is >0.05 thus not statistically significant

MODEL



- demand for bikes are slightly less likely on humid days

'WINDSPEED'

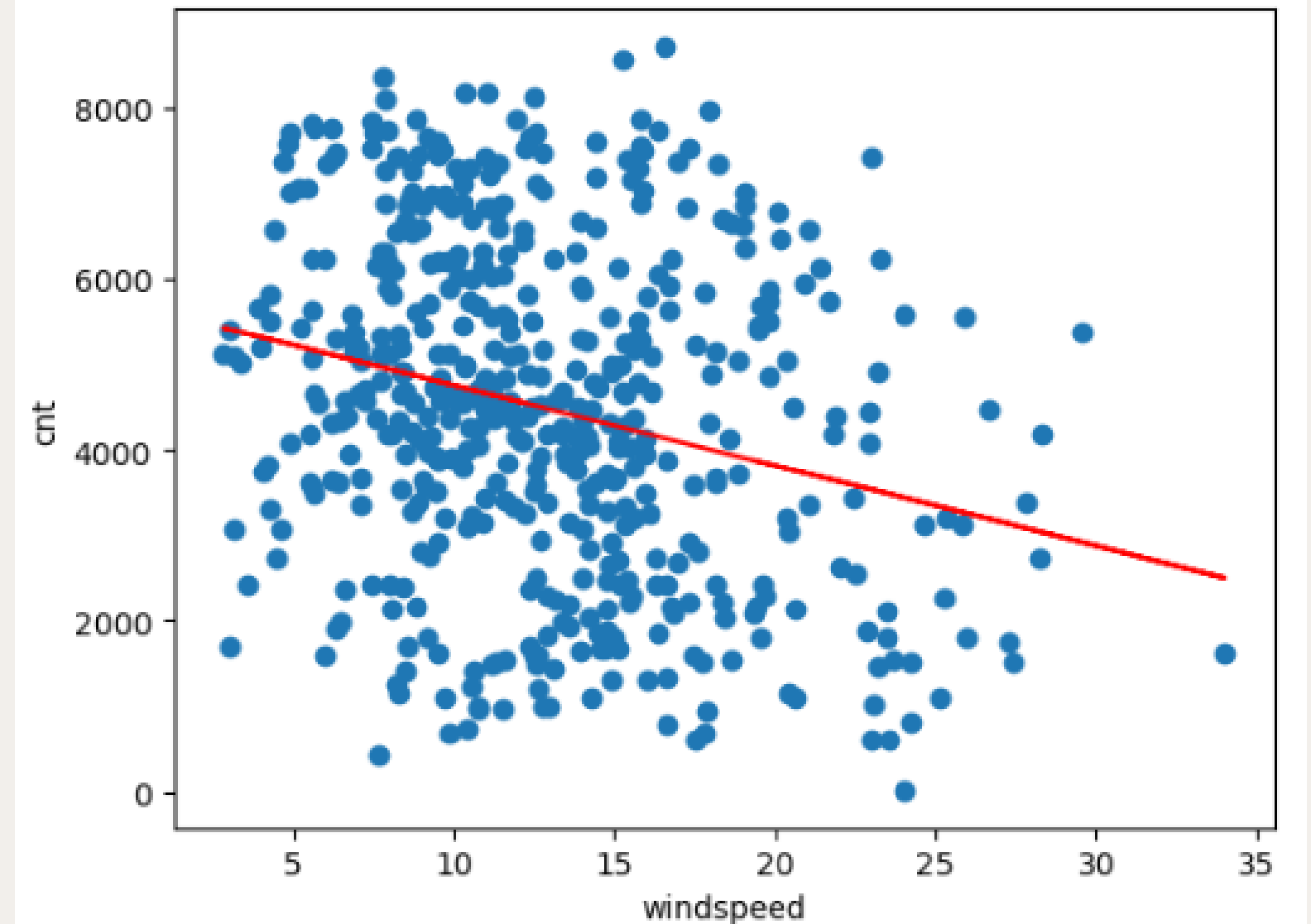
OLS Regression Results

```
=====
Dep. Variable:          cnt    R-squared:                0.064
Model:                  OLS    Adj. R-squared:            0.063
Method:                 Least Squares    F-statistic:        34.97
Date:                   Wed, 10 May 2023    Prob (F-statistic):    6.14e-09
Time:                   00:27:50    Log-Likelihood:        -4570.3
No. Observations:       510    AIC:                  9145.
Df Residuals:           508    BIC:                  9153.
Df Model:                1
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	5687.7099	219.704	25.888	0.000	5256.069	6119.350
windspeed	-93.6246	15.831	-5.914	0.000	-124.728	-62.522

```
=====
Omnibus:                33.829    Durbin-Watson:          1.872
Prob(Omnibus):           0.000    Jarque-Bera (JB):       12.465
Skew:                    -0.002    Prob(JB):               0.00196
Kurtosis:                2.234    Cond. No.               36.6
=====
```

MODEL



- demand lowers as windspeed increases

'SEASON'

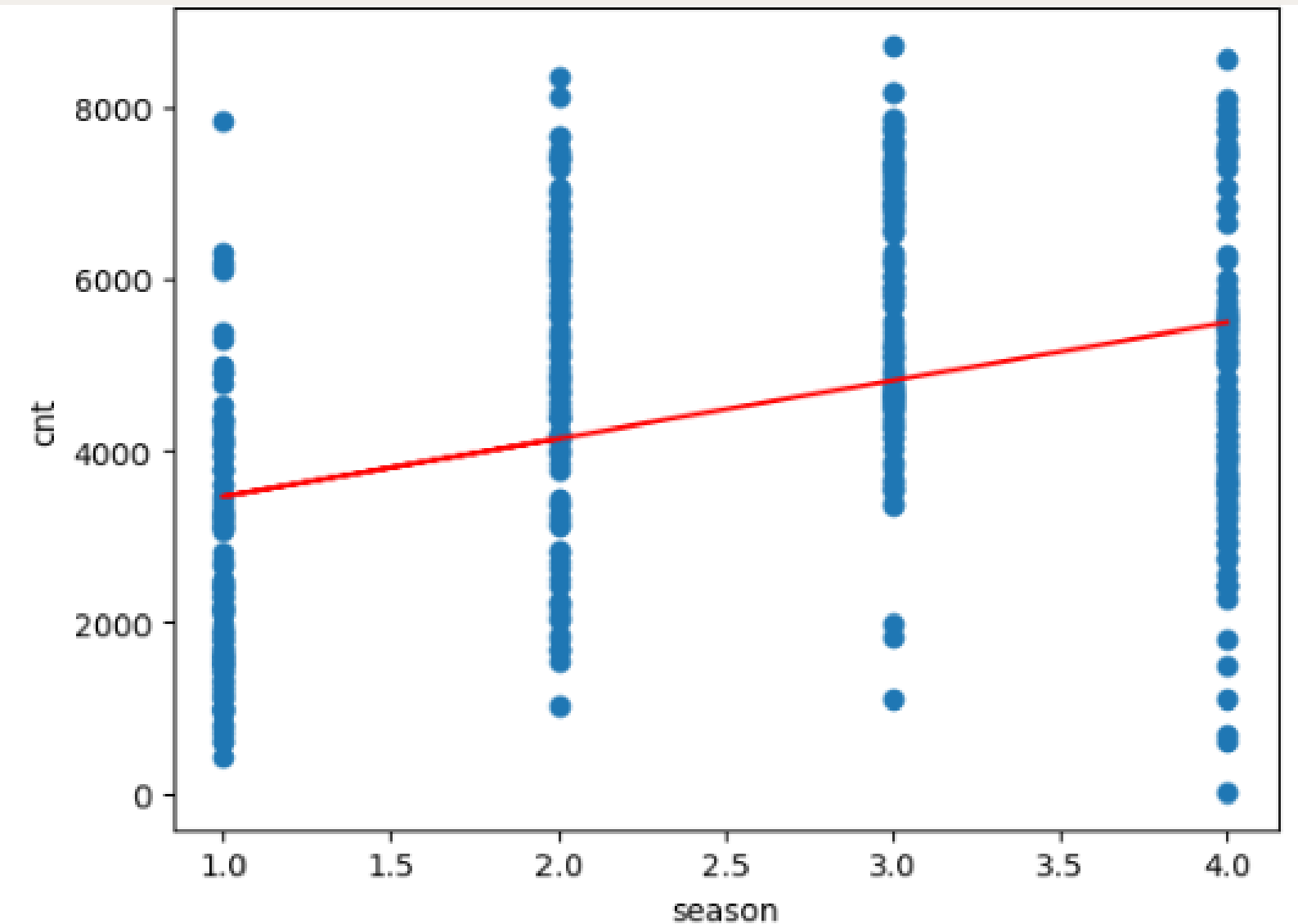
OLS Regression Results

```
=====
Dep. Variable:          cnt      R-squared:          0.148
Model:                  OLS      Adj. R-squared:      0.146
Method:                 Least Squares      F-statistic:      88.18
Date:                   Wed, 10 May 2023      Prob (F-statistic):  2.00e-19
Time:                   00:27:49      Log-Likelihood:     -4546.5
No. Observations:       510      AIC:              9097.
Df Residuals:           508      BIC:              9105.
Df Model:                1
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	2786.4475	197.862	14.083	0.000	2397.720	3175.175
season	675.2078	71.902	9.391	0.000	533.946	816.470

```
=====
Omnibus:                20.738      Durbin-Watson:          1.882
Prob(Omnibus):           0.000      Jarque-Bera (JB):        9.503
Skew:                    0.052      Prob(JB):                0.00864
Kurtosis:                2.340      Cond. No.:               7.59
=====
```

MODEL



- demand is non-fluctuant through the seasons

MULTIPLE LINEAR REGRESSION ANALYSIS

OLS Regression Results

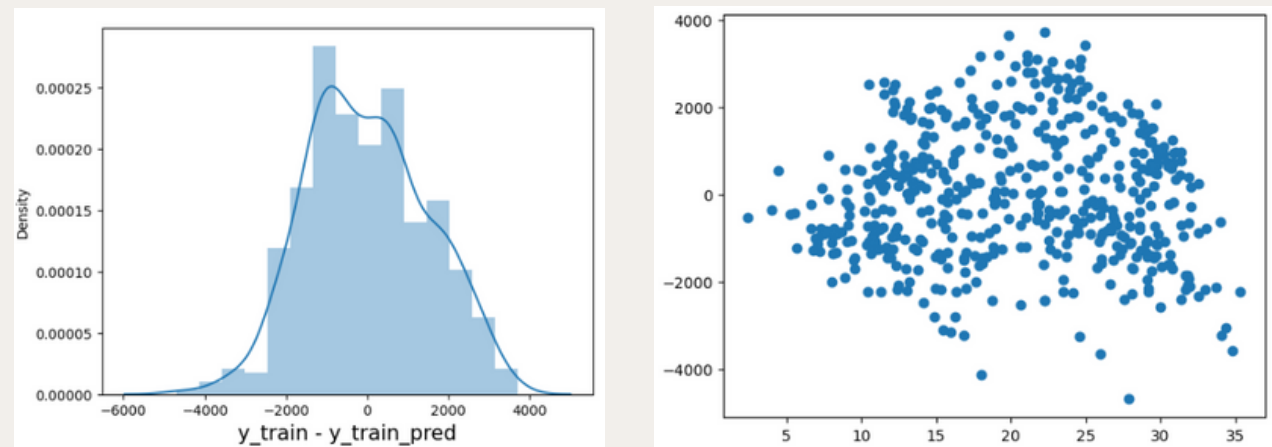
```
=====
Dep. Variable:          cnt      R-squared:          0.528
Model:                  OLS      Adj. R-squared:       0.524
Method:                 Least Squares      F-statistic:       112.9
Date:                  Wed, 17 May 2023      Prob (F-statistic):   6.81e-80
Time:                  22:14:13      Log-Likelihood:      -4395.7
No. Observations:      510      AIC:                8803.
Df Residuals:          504      BIC:                8829.
Df Model:              5
Covariance Type:       nonrobust
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const      3286.2233     394.035      8.340      0.000     2512.069     4060.377
hum        -18.8501       5.637     -3.344      0.001     -29.926     -7.775
weathersit  -587.2719     139.133     -4.221      0.000     -860.623    -313.921
windspeed  -50.1911      12.183     -4.120      0.000     -74.126     -26.256
season      371.8064      58.390      6.368      0.000      257.088     486.524
temp       145.3117       8.759     16.589      0.000      128.102     162.521
=====
Omnibus:          6.998      Durbin-Watson:       2.021
Prob(Omnibus):    0.030      Jarque-Bera (JB):     4.846
Skew:            0.090      Prob(JB):             0.0886
Kurtosis:        2.557      Cond. No.             455.
=====
```

R-Squared Value is 0.528. 52.8% of the variance in demand is explained by the independent variables.

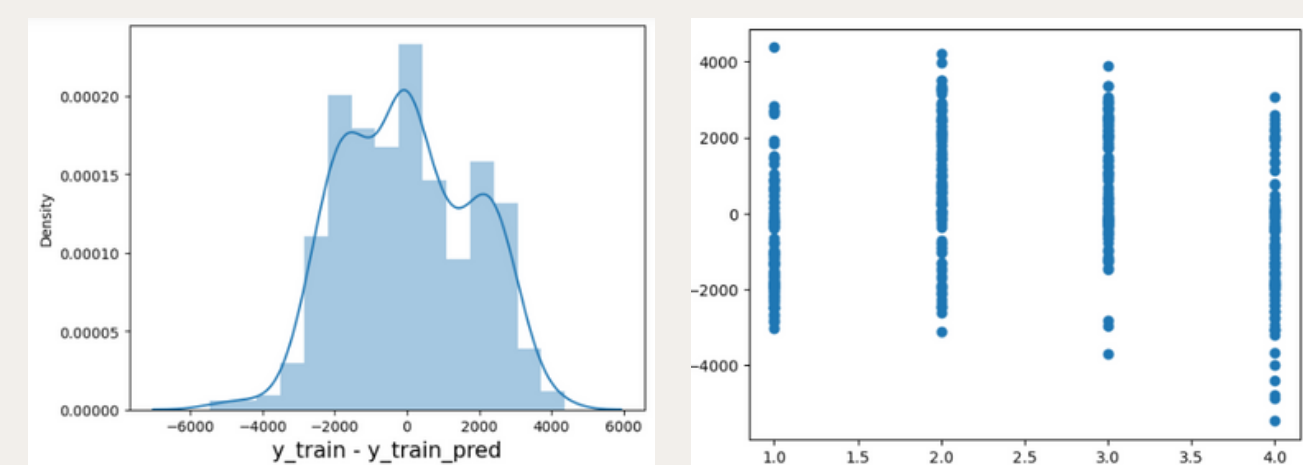
All p values indicate all variables together are statistically significant.

RESIDUAL ANALYSIS

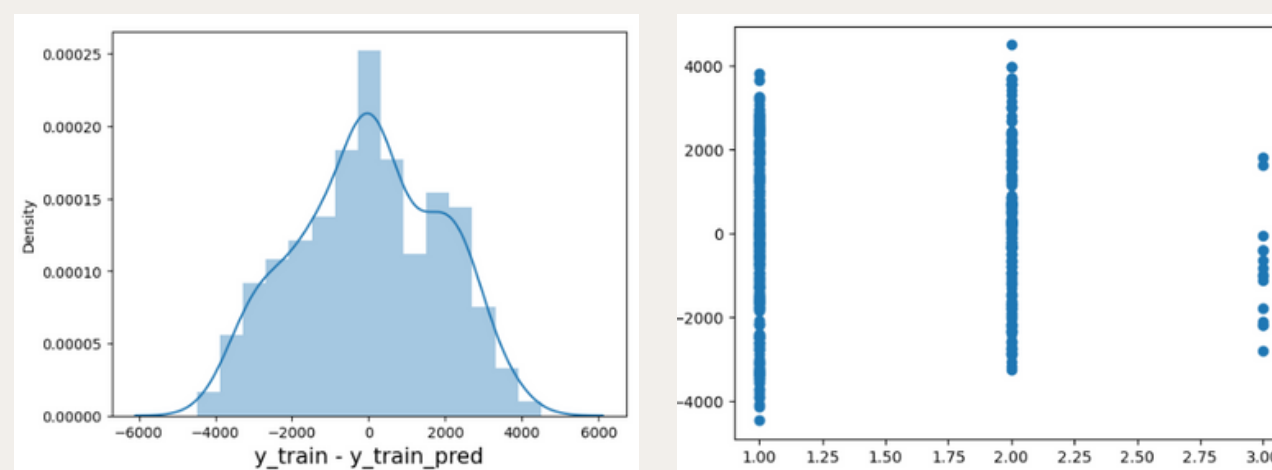
'temp'



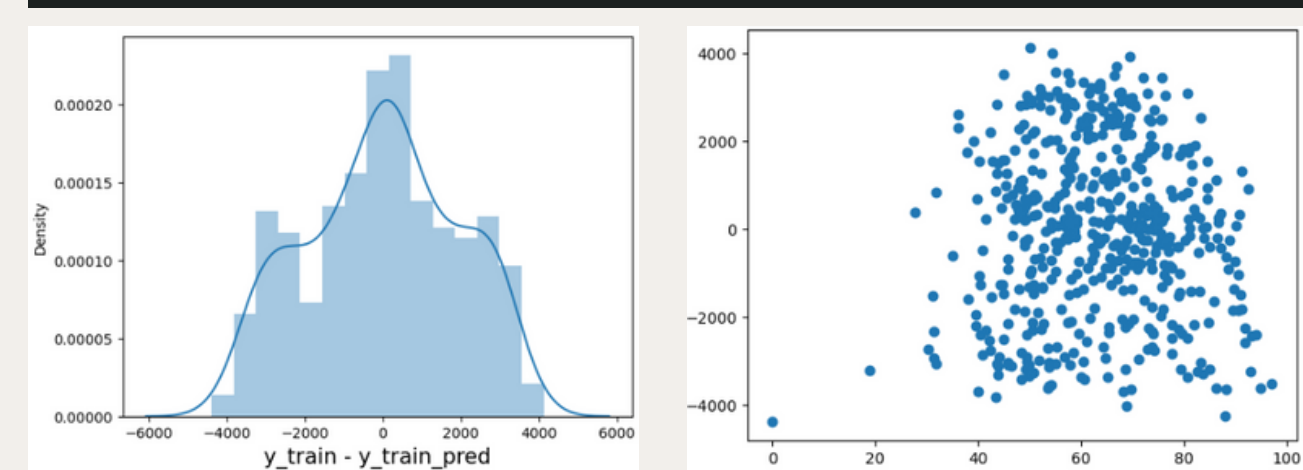
'season'



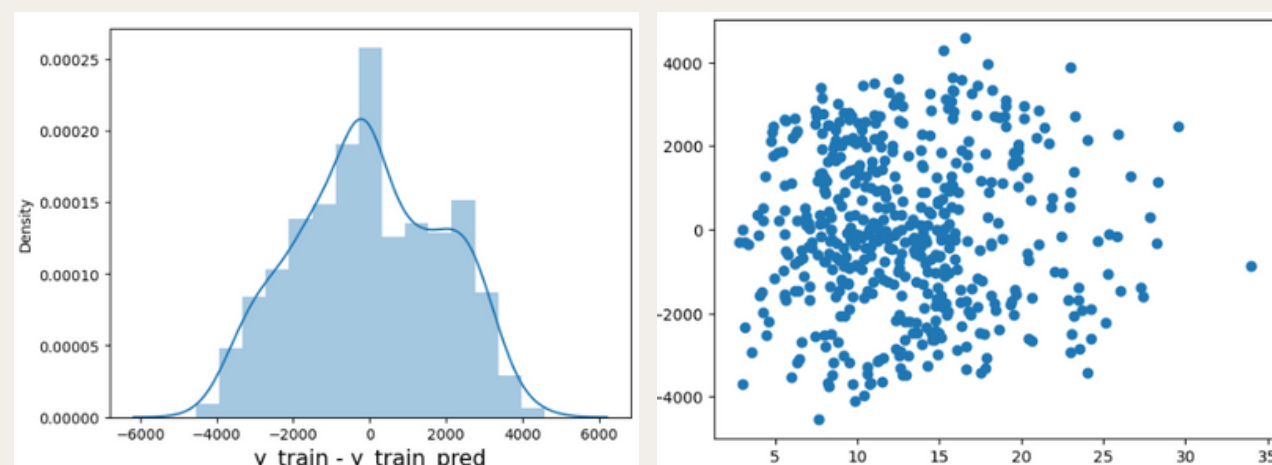
'weathersit'



'hum'



'windspeed'



INFERENCE

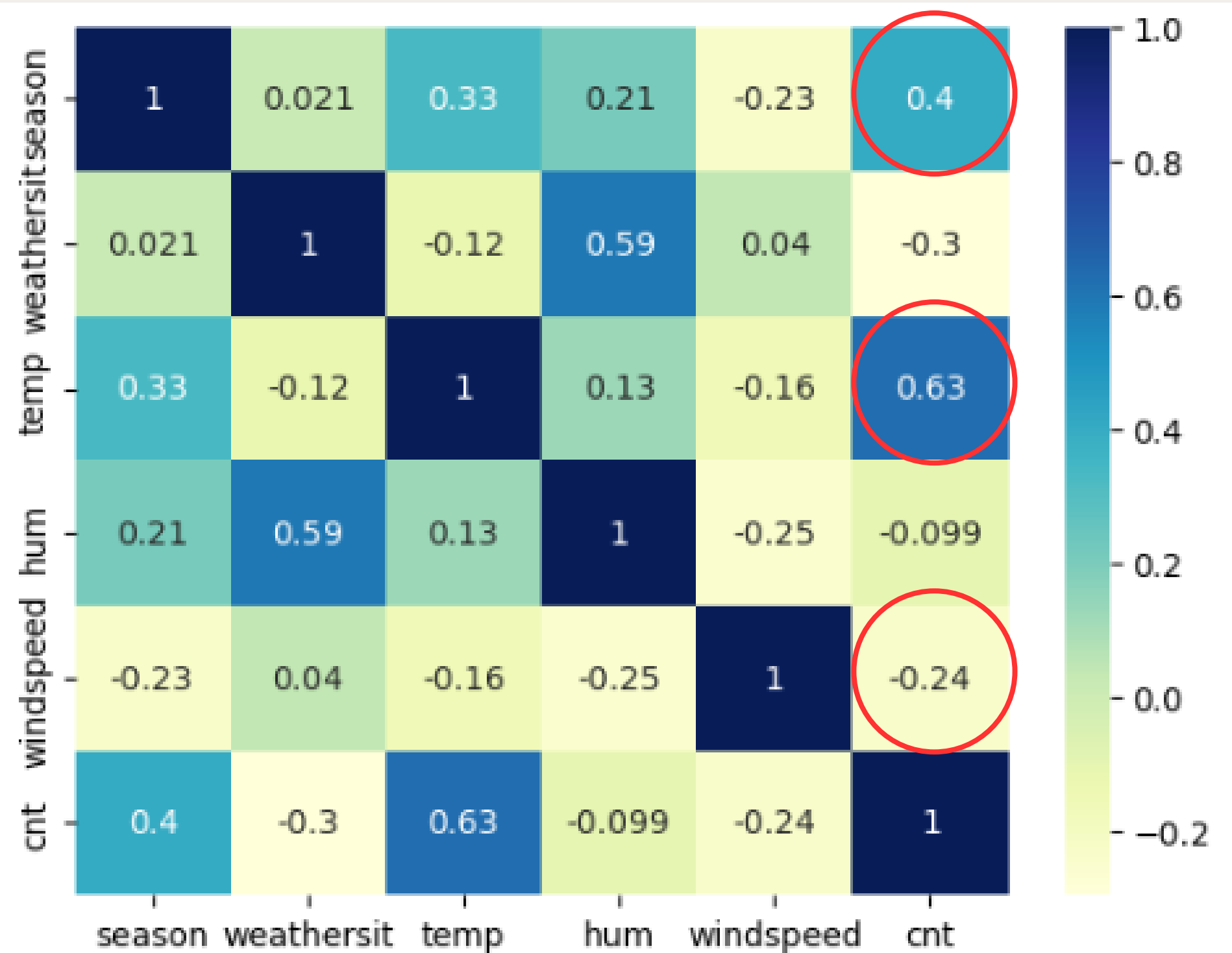
- all variables are normally distributed except 'season' and 'temp'
- all variables has significant amount of residuals

CORRELATION HEATMAP

INFERENCE

The top 3 variables with the highest correlation to the target variable are

1. **temp** (temperature)
2. **season**
3. **windspeed**



OVERVIEW

Based on the findings, we can conclude that the outside temperature is the variable with the highest predictor for demand of bikes. If outside temperatures are hot, people are more likely to rent out bikes. Furthermore, the demand is significantly affected by the season as well. The company tend to have higher demand during hot seasons compared to cold seasons. Lastly, outside factors such as the weather and wind speed also affects the demand of bikes. Bikes are more likely to be rented out on clear less windy days. Boombikes have to take account all these variables in order to predict the their future demands.