# Linear

June 10, 2024

[1]:
```
# It may take several minutes to install those libraries in Watson Studio
install.packages("rlang")
```

```
Updating HTML index of packages in '.Library'
Making 'packages.html' … done
```

[2]:
```
# It may take several minutes to install those libraries in Watson Studio
library("tidymodels")
library("tidyverse")
library("stringr")
```

```
Warning message:
"replacing previous import ‘lifecycle::last_warnings’ by ‘rlang::last_warnings’
when loading ‘tibble’"Warning message:
"replacing previous import ‘ellipsis::check_dots_unnamed’ by
‘rlang::check_dots_unnamed’ when loading ‘tibble’"Warning message:
"replacing previous import ‘ellipsis::check_dots_used’ by
‘rlang::check_dots_used’ when loading ‘tibble’"Warning message:
"replacing previous import ‘ellipsis::check_dots_empty’ by
‘rlang::check_dots_empty’ when loading ‘tibble’"  Attaching packages
                     tidymodels 0.1.0
 broom     0.5.6       recipes   0.1.12
 dials     0.0.6       rsample   0.0.5
 dplyr     0.8.5       tibble    3.0.1
 ggplot2   3.3.0       tune      0.1.0
 infer     0.5.1       workflows 0.1.1
 parsnip   0.1.0       yardstick 0.0.6
 purrr     0.3.4
  Conflicts                         tidymodels_conflicts()
 purrr::discard()  masks scales::discard()
 dplyr::filter()   masks stats::filter()
 dplyr::lag()      masks stats::lag()
 ggplot2::margin() masks dials::margin()
 recipes::step()   masks stats::step()
  Attaching packages                          tidyverse 1.3.0
 readr   1.3.1      forcats 0.5.0
 stringr 1.4.0
  Conflicts                         tidyverse_conflicts()
```

```
readr::col_factor() masks scales::col_factor()
purrr::discard()    masks scales::discard()
dplyr::filter()     masks stats::filter()
stringr::fixed()    masks recipes::fixed()
dplyr::lag()        masks stats::lag()
ggplot2::margin()   masks dials::margin()
readr::spec()       masks yardstick::spec()
```

[3]:
```
# Dataset URL
dataset_url <- "https://cf-courses-data.s3.us.cloud-object-storage.appdomain.
 ↪cloud/IBMDeveloperSkillsNetwork-RP0321EN-SkillsNetwork/labs/datasets/
 ↪seoul_bike_sharing_converted_normalized.csv"
bike_sharing_df <- read_csv(dataset_url)
spec(bike_sharing_df)
```

```
Parsed with column specification:
cols(
  .default = col_double(),
  DATE = col_character(),
  FUNCTIONING_DAY = col_character()
)
See spec(…) for full column specifications.
```

```
cols(
  DATE = col_character(),
  RENTED_BIKE_COUNT = col_double(),
  TEMPERATURE = col_double(),
  HUMIDITY = col_double(),
  WIND_SPEED = col_double(),
  VISIBILITY = col_double(),
  DEW_POINT_TEMPERATURE = col_double(),
  SOLAR_RADIATION = col_double(),
  RAINFALL = col_double(),
  SNOWFALL = col_double(),
  FUNCTIONING_DAY = col_character(),
  `0` = col_double(),
  `1` = col_double(),
  `10` = col_double(),
  `11` = col_double(),
  `12` = col_double(),
  `13` = col_double(),
  `14` = col_double(),
  `15` = col_double(),
  `16` = col_double(),
  `17` = col_double(),
  `18` = col_double(),
  `19` = col_double(),
  `2` = col_double(),
```

```
  `20` = col_double(),
  `21` = col_double(),
  `22` = col_double(),
  `23` = col_double(),
  `3` = col_double(),
  `4` = col_double(),
  `5` = col_double(),
  `6` = col_double(),
  `7` = col_double(),
  `8` = col_double(),
  `9` = col_double(),
  AUTUMN = col_double(),
  SPRING = col_double(),
  SUMMER = col_double(),
  WINTER = col_double(),
  HOLIDAY = col_double(),
  NO_HOLIDAY = col_double()
)
```

[4]:
```
bike_sharing_df <- bike_sharing_df %>%
                    select(-DATE, -FUNCTIONING_DAY)
```

[5]:
```
# Use the `initial_split()`, `training()`, and `testing()` functions to split␣
 ↪the dataset
# With seed 1234
set.seed(1234)
# prop = 3/4
bike_sharing_split <- initial_split(bike_sharing_df, prop = 0.75)
# train_data
train_data <- training(bike_sharing_split)
# test_data
test_data <- testing(bike_sharing_split)
```

[6]:
```
##TASK: Build a linear regression model using weather variables only␣
 ↪lm_model_weather
### Use `linear_reg()` with engine `lm` and mode `regression`
lm_model_weather <- linear_reg( mode ="regression") %>%
# Set engine
set_engine(engine = "lm")
# Print the linear function
lm_model_weather
```

```
Linear Regression Model Specification (regression)

Computational engine: lm
```

[7]:
```
train_fit <- lm_model_weather %>%
fit(RENTED_BIKE_COUNT ~ TEMPERATURE + HUMIDITY + WIND_SPEED + VISIBILITY +␣
  ↪DEW_POINT_TEMPERATURE + SOLAR_RADIATION + RAINFALL + SNOWFALL, data =␣
  ↪train_data)
train_fit
```

```
parsnip model object

Fit time:  7ms

Call:
stats::lm(formula = formula, data = data)

Coefficients:
          (Intercept)            TEMPERATURE               HUMIDITY
              147.647               2452.112               -895.830
         WIND_SPEED             VISIBILITY  DEW_POINT_TEMPERATURE
              402.183                  5.356               -368.982
      SOLAR_RADIATION               RAINFALL               SNOWFALL
             -435.703              -1771.467                354.761
```

[8]:
```
##TASK: Model evaluation and identification of important variables
# Use predict() function to generate test results for `lm_model_weather` and␣
  ↪`lm_model_all`
train_results <- train_fit %>%
# Make the predictions and save the predicted values
predict(new_data = train_data) %>%

# Create a new column to save the true values
mutate(truth = train_data$RENTED_BIKE_COUNT)
head(train_results)
```

A tibble: 6 × 2

| .pred <dbl> | truth <dbl> |
|---|---|
| 391.5112 | 254 |
| 293.4206 | 204 |
| 250.5649 | 107 |
| 309.0461 | 100 |
| 228.2566 | 460 |
| 241.6479 | 930 |

[9]:
```
test_results <- train_fit %>%
# Make the predictions and save the predicted values
predict(new_data = test_data) %>%
# Create a new column to save the true values
mutate(truth = test_data$RENTED_BIKE_COUNT)
```

```
head(test_results)
```

A tibble: 6 × 2

| .pred <dbl> | truth <dbl> |
|---|---|
| 274.3531 | 173 |
| 378.1757 | 78 |
| 312.9917 | 181 |
| 336.7766 | 490 |
| 631.8560 | 449 |
| 638.0489 | 451 |

[10]:
```
lm_model_all <- linear_reg( mode ="regression") %>%
# Set engine
set_engine(engine = "lm")
# Print the linear function
lm_model_all
```

```
Linear Regression Model Specification (regression)

Computational engine: lm
```

[11]:
```
train_fit2 <- lm_model_all %>%
fit(RENTED_BIKE_COUNT ~ ., data = train_data)
train_fit2
```

```
parsnip model object

Fit time:  13ms

Call:
stats::lm(formula = formula, data = data)

Coefficients:
          (Intercept)           TEMPERATURE              HUMIDITY
              216.584               810.604              -920.587
           WIND_SPEED            VISIBILITY  DEW_POINT_TEMPERATURE
               -9.313                24.368               632.384
      SOLAR_RADIATION              RAINFALL              SNOWFALL
              249.752             -1982.940               232.451
                  `0`                   `1`                   `10`
              -12.500              -125.701              -222.233
                 `11`                  `12`                   `13`
             -231.428              -194.605              -189.528
                 `14`                  `15`                   `16`
             -183.060              -103.127                43.598
                 `17`                  `18`                   `19`
              306.732               782.842               519.456
```

| | | |
|---|---|---|
| `2` | `20` | `21` |
| -259.946 | 394.635 | 430.185 |
| `22` | `23` | `3` |
| 322.664 | 90.181 | -327.245 |
| `4` | `5` | `6` |
| -389.558 | -379.055 | -216.689 |
| `7` | `8` | `9` |
| 110.492 | 502.354 | NA |
| AUTUMN | SPRING | SUMMER |
| 357.978 | 194.421 | 172.901 |
| WINTER | HOLIDAY | NO_HOLIDAY |
| NA | -129.167 | NA |

[12]:
```
train_results2 <- train_fit2 %>%
# Make the predictions and save the predicted values
predict(new_data = train_data) %>%

# Create a new column to save the true values
mutate(truth = train_data$RENTED_BIKE_COUNT)
head(train_results2)
```

```
Warning message in predict.lm(object = object$fit, newdata = new_data, type =
"response"):
"prediction from a rank-deficient fit may be misleading"
```

A tibble: 6 × 2

| .pred <dbl> | truth <dbl> |
|---|---|
| 198.90633 | 254 |
| 73.82145 | 204 |
| -156.55532 | 107 |
| -195.80875 | 100 |
| 264.36386 | 460 |
| 657.77254 | 930 |

[13]:
```
test_results2 <- train_fit2 %>%
# Make the predictions and save the predicted values
predict(new_data = test_data) %>%
# Create a new column to save the true values
mutate(truth = test_data$RENTED_BIKE_COUNT)
head(test_results2)
```

```
Warning message in predict.lm(object = object$fit, newdata = new_data, type =
"response"):
"prediction from a rank-deficient fit may be misleading"
```

A tibble: 6 × 2

| .pred | truth |
|---|---|
| <dbl> | <dbl> |
| -78.24847 | 173 |
| -191.16239 | 78 |
| -25.99046 | 181 |
| 251.97376 | 490 |
| 330.23653 | 449 |
| 347.24700 | 451 |

[14]:
```
rsq(test_results, truth = truth,
estimate = .pred)

rsq(test_results2, truth = truth,
estimate = .pred)
```

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rsq | standard | 0.4458692 |

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rsq | standard | 0.6592615 |

[15]:
```
rmse(test_results, truth = truth,
estimate = .pred)

rmse(test_results2, truth = truth,
estimate = .pred)
```

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rmse | standard | 470.4436 |

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rmse | standard | 369.1709 |

[16]:
```
## summary(lm_model_all$fit)
lm_model_all$coefficients
```

```
NULL
```

[17]:
```
lm_model_weather$coefficients
```

```
NULL
```

[18]:
```
train_fit2  %>%
  tidy() %>%
  arrange(desc(abs(estimate)))
```

A tibble: 36 × 5

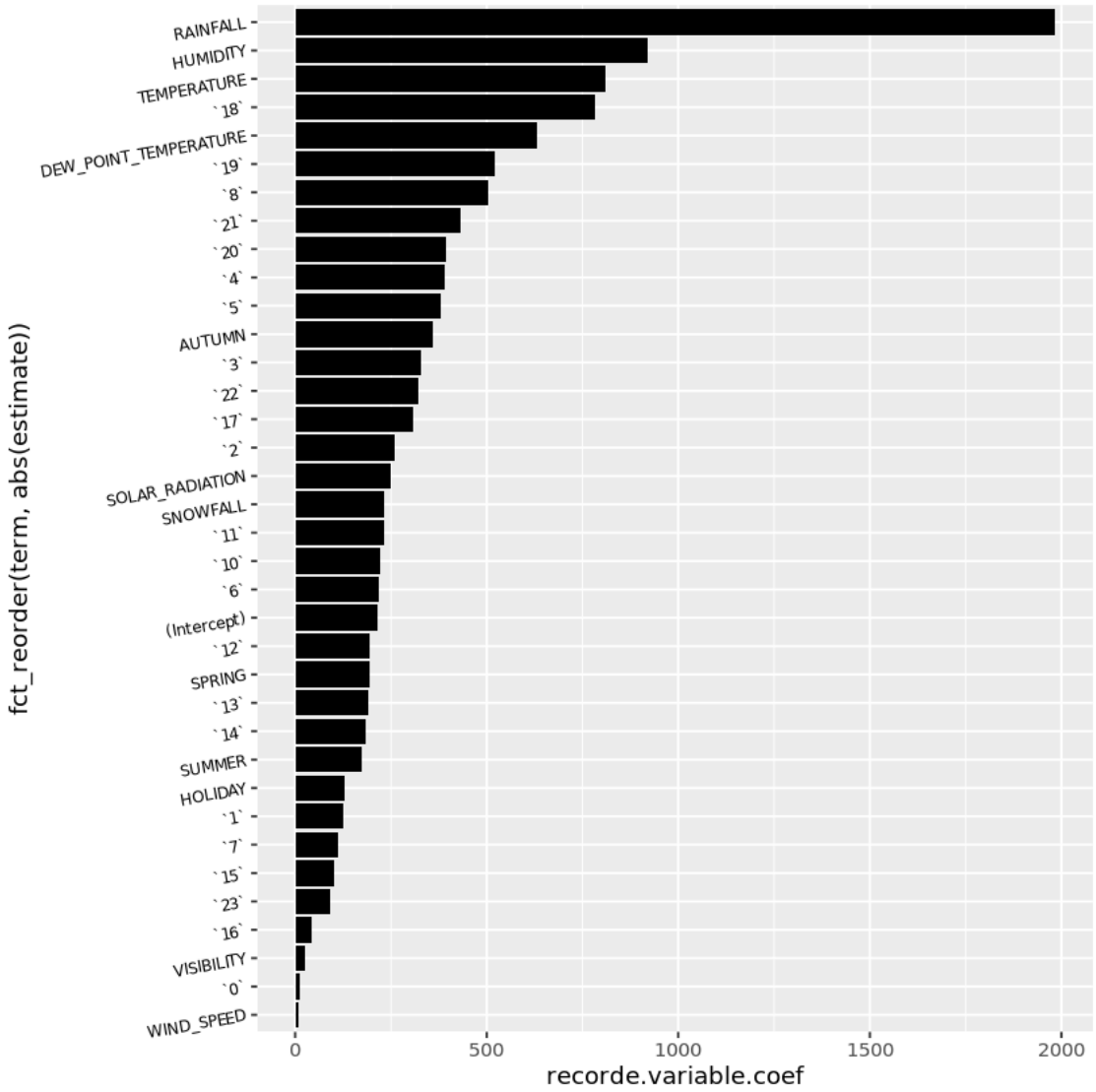
| term<br><chr> | estimate<br><dbl> | std.error<br><dbl> | statistic<br><dbl> | p.value<br><dbl> |
|---|---|---|---|---|
| RAINFALL | -1982.940423 | 143.64226 | -13.8047146 | 9.922612e-43 |
| HUMIDITY | -920.586909 | 97.24275 | -9.4668947 | 3.978466e-21 |
| TEMPERATURE | 810.603987 | 209.03372 | 3.8778623 | 1.064486e-04 |
| ‘18‘ | 782.841507 | 34.51207 | 22.6831196 | 1.428163e-109 |
| DEW_POINT_TEMPERATURE | 632.384298 | 217.47791 | 2.9078093 | 3.652385e-03 |
| ‘19‘ | 519.455508 | 34.54060 | 15.0389836 | 3.001427e-50 |
| ‘8‘ | 502.353808 | 33.50621 | 14.9928559 | 5.875496e-50 |
| ‘21‘ | 430.185215 | 34.33684 | 12.5283880 | 1.380106e-35 |
| ‘20‘ | 394.634676 | 34.72886 | 11.3633072 | 1.234052e-29 |
| ‘4‘ | -389.558392 | 34.03286 | -11.4465363 | 4.834550e-30 |
| ‘5‘ | -379.055209 | 33.81771 | -11.2087790 | 6.910457e-29 |
| AUTUMN | 357.978265 | 19.99659 | 17.9019693 | 5.967468e-70 |
| ‘3‘ | -327.244955 | 33.88966 | -9.6561893 | 6.555260e-22 |
| ‘22‘ | 322.663949 | 33.93497 | 9.5083031 | 2.689464e-21 |
| ‘17‘ | 306.732435 | 34.27155 | 8.9500589 | 4.599800e-19 |
| ‘2‘ | -259.946174 | 34.57530 | -7.5182614 | 6.322812e-14 |
| SOLAR_RADIATION | 249.752454 | 41.34675 | 6.0404379 | 1.624462e-09 |
| SNOWFALL | 232.450790 | 104.60901 | 2.2220915 | 2.631243e-02 |
| ‘11‘ | -231.428411 | 33.98053 | -6.8106172 | 1.061643e-11 |
| ‘10‘ | -222.233262 | 33.31558 | -6.6705509 | 2.764896e-11 |
| ‘6‘ | -216.689105 | 34.16556 | -6.3423250 | 2.419940e-10 |
| (Intercept) | 216.584059 | 50.67322 | 4.2741329 | 1.947061e-05 |
| ‘12‘ | -194.605041 | 34.55857 | -5.6311662 | 1.867030e-08 |
| SPRING | 194.421437 | 19.21791 | 10.1166801 | 7.080230e-24 |
| ‘13‘ | -189.528426 | 35.19884 | -5.3845072 | 7.526455e-08 |
| ‘14‘ | -183.060291 | 35.24584 | -5.1938130 | 2.124823e-07 |
| SUMMER | 172.901302 | 29.04270 | 5.9533473 | 2.768259e-09 |
| HOLIDAY | -129.167078 | 22.44739 | -5.7542142 | 9.112956e-09 |
| ‘1‘ | -125.701499 | 34.57857 | -3.6352430 | 2.799291e-04 |
| ‘7‘ | 110.491959 | 33.93040 | 3.2564298 | 1.134185e-03 |
| ‘15‘ | -103.126885 | 35.09640 | -2.9383890 | 3.311170e-03 |
| ‘23‘ | 90.180803 | 33.97406 | 2.6544016 | 7.964847e-03 |
| ‘16‘ | 43.597895 | 34.23026 | 1.2736652 | 2.028290e-01 |
| VISIBILITY | 24.368049 | 20.23481 | 1.2042639 | 2.285328e-01 |
| ‘0‘ | -12.499804 | 34.01006 | -0.3675326 | 7.132341e-01 |
| WIND_SPEED | -9.313319 | 40.26779 | -0.2312846 | 8.171012e-01 |

[19]:
```
train_fit2 %>%
  tidy() %>%
  filter(!is.na(estimate)) %>%
  ggplot(aes(x = fct_reorder(term, abs(estimate)), y = abs(estimate))) +
  geom_bar(stat = "identity", fill = "black") +
  coord_flip() +
  theme(axis.text.y = element_text(angle = 10, colour = "black", size = 7)) +
  ylab("recorde.variable.coef")
```

```
  xlab("Coef")
```

```
$x
[1] "Coef"

attr(,"class")
[1] "labels"
```




[20]:
```
# Dataset URL
dataset_url <- "https://cf-courses-data.s3.us.cloud-object-storage.appdomain.
  ↪cloud/IBMDeveloperSkillsNetwork-RP0321EN-SkillsNetwork/labs/datasets/
  ↪seoul_bike_sharing_converted.csv"
bike_sharing_df <- read_csv(dataset_url)
```

```
spec(bike_sharing_df)
```

```
Parsed with column specification:
cols(
  .default = col_double(),
  DATE = col_character(),
  FUNCTIONING_DAY = col_character()
)
See spec(…) for full column specifications.
```

```
cols(
  DATE = col_character(),
  RENTED_BIKE_COUNT = col_double(),
  TEMPERATURE = col_double(),
  HUMIDITY = col_double(),
  WIND_SPEED = col_double(),
  VISIBILITY = col_double(),
  DEW_POINT_TEMPERATURE = col_double(),
  SOLAR_RADIATION = col_double(),
  RAINFALL = col_double(),
  SNOWFALL = col_double(),
  FUNCTIONING_DAY = col_character(),
  `0` = col_double(),
  `1` = col_double(),
  `10` = col_double(),
  `11` = col_double(),
  `12` = col_double(),
  `13` = col_double(),
  `14` = col_double(),
  `15` = col_double(),
  `16` = col_double(),
  `17` = col_double(),
  `18` = col_double(),
  `19` = col_double(),
  `2` = col_double(),
  `20` = col_double(),
  `21` = col_double(),
  `22` = col_double(),
  `23` = col_double(),
  `3` = col_double(),
  `4` = col_double(),
  `5` = col_double(),
  `6` = col_double(),
  `7` = col_double(),
  `8` = col_double(),
  `9` = col_double(),
  AUTUMN = col_double(),
  SPRING = col_double(),
```

```
  SUMMER = col_double(),
  WINTER = col_double(),
  HOLIDAY = col_double(),
  NO_HOLIDAY = col_double()
)
```

[21]:
```
set.seed(1234)
# prop = 3/4
bike_sharing_split2 <- initial_split(bike_sharing_df, prop = 0.75)
# train_data
train_data1 <- training(bike_sharing_split2)
# test_data
test_data2 <- testing(bike_sharing_split2)
```

[22]:
```
lm_model_weather <- linear_reg( mode ="regression") %>%
# Set engine
set_engine(engine = "lm")
# Print the linear function
lm_model_weather
```

```
Linear Regression Model Specification (regression)

Computational engine: lm
```

[23]:
```
train_fit <- lm_model_weather %>%
fit(RENTED_BIKE_COUNT ~ TEMPERATURE + HUMIDITY + WIND_SPEED + VISIBILITY +␣
  ↪DEW_POINT_TEMPERATURE + SOLAR_RADIATION + RAINFALL + SNOWFALL, data =␣
  ↪train_data)
train_fit
```

```
parsnip model object

Fit time:  4ms

Call:
stats::lm(formula = formula, data = data)

Coefficients:
          (Intercept)           TEMPERATURE              HUMIDITY
              147.647              2452.112              -895.830
         WIND_SPEED            VISIBILITY  DEW_POINT_TEMPERATURE
              402.183                 5.356              -368.982
      SOLAR_RADIATION              RAINFALL              SNOWFALL
             -435.703             -1771.467               354.761
```

[24]:
```
test_results <- train_fit %>%
# Make the predictions and save the predicted values
predict(new_data = test_data) %>%
# Create a new column to save the true values
mutate(truth = test_data$RENTED_BIKE_COUNT)
head(test_results)
```

A tibble: 6 × 2

| .pred <dbl> | truth <dbl> |
|---|---|
| 274.3531 | 173 |
| 378.1757 | 78 |
| 312.9917 | 181 |
| 336.7766 | 490 |
| 631.8560 | 449 |
| 638.0489 | 451 |

[25]:
```
lm_model_all <- linear_reg( mode ="regression") %>%
# Set engine
set_engine(engine = "lm")
# Print the linear function
lm_model_all
```

```
Linear Regression Model Specification (regression)

Computational engine: lm
```

[26]:
```
train_fit2 <- lm_model_all %>%
fit(RENTED_BIKE_COUNT ~ ., data = train_data)
train_fit2
```

```
parsnip model object

Fit time:  18ms

Call:
stats::lm(formula = formula, data = data)

Coefficients:
          (Intercept)           TEMPERATURE              HUMIDITY
              216.584               810.604              -920.587
          WIND_SPEED            VISIBILITY  DEW_POINT_TEMPERATURE
               -9.313                24.368               632.384
      SOLAR_RADIATION              RAINFALL              SNOWFALL
              249.752             -1982.940               232.451
                  `0`                   `1`                  `10`
              -12.500              -125.701              -222.233
                 `11`                  `12`                  `13`
```

```
-231.428        -194.605        -189.528
    `14`            `15`            `16`
-183.060        -103.127          43.598
    `17`            `18`            `19`
 306.732         782.842         519.456
     `2`            `20`            `21`
-259.946         394.635         430.185
    `22`            `23`             `3`
 322.664          90.181        -327.245
     `4`             `5`             `6`
-389.558        -379.055        -216.689
     `7`             `8`             `9`
 110.492         502.354              NA
  AUTUMN          SPRING          SUMMER
 357.978         194.421         172.901
  WINTER          HOLIDAY      NO_HOLIDAY
      NA        -129.167              NA
```

[27]:
```
test_results2 <- train_fit2 %>%
# Make the predictions and save the predicted values
predict(new_data = test_data) %>%
# Create a new column to save the true values
mutate(truth = test_data$RENTED_BIKE_COUNT)
head(test_results2)
```

```
Warning message in predict.lm(object = object$fit, newdata = new_data, type =
"response"):
"prediction from a rank-deficient fit may be misleading"
```

A tibble: 6 × 2

| .pred <dbl> | truth <dbl> |
|---|---|
| -78.24847 | 173 |
| -191.16239 | 78 |
| -25.99046 | 181 |
| 251.97376 | 490 |
| 330.23653 | 449 |
| 347.24700 | 451 |

[28]:
```
rsq(test_results, truth = truth,
estimate = .pred)
rsq(test_results2, truth = truth,
estimate = .pred)

rmse(test_results, truth = truth,
estimate = .pred)
rmse(test_results2, truth = truth,
estimate = .pred)
```

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rsq | standard | 0.4458692 |

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rsq | standard | 0.6592615 |

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rmse | standard | 470.4436 |

A tibble: 1 × 3

| .metric | .estimator | .estimate |
|---|---|---|
| <chr> | <chr> | <dbl> |
| rmse | standard | 369.1709 |

[29]:
```
lm_model_all$fit$coefficients
```

```
NULL
```

[30]:
```
train_fit2  %>%
  tidy() %>%
  arrange(desc(abs(estimate)))
```

A tibble: 36 × 5

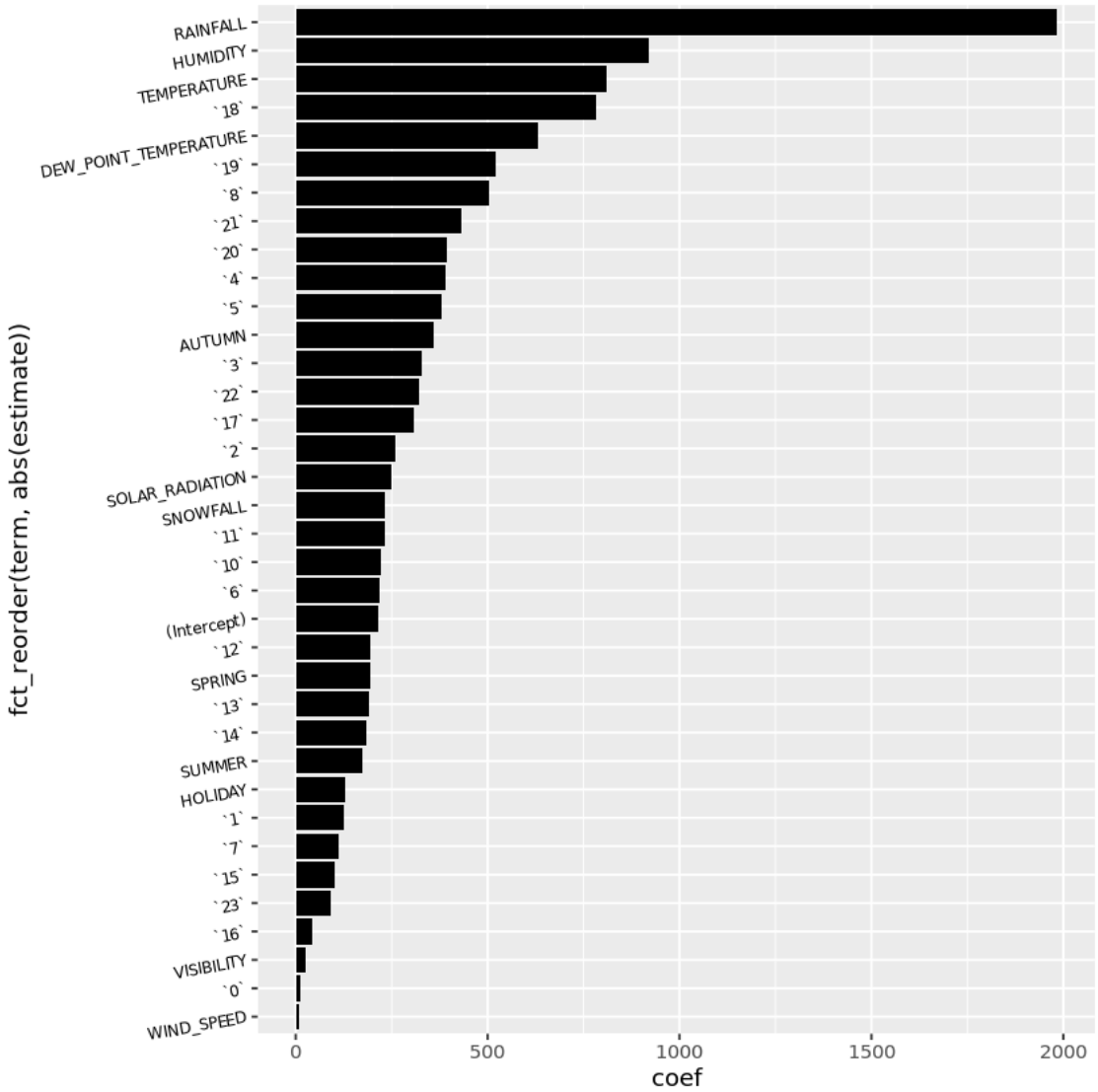
| term<br><chr> | estimate<br><dbl> | std.error<br><dbl> | statistic<br><dbl> | p.value<br><dbl> |
|---|---|---|---|---|
| RAINFALL | -1982.940423 | 143.64226 | -13.8047146 | 9.922612e-43 |
| HUMIDITY | -920.586909 | 97.24275 | -9.4668947 | 3.978466e-21 |
| TEMPERATURE | 810.603987 | 209.03372 | 3.8778623 | 1.064486e-04 |
| ‘18‘ | 782.841507 | 34.51207 | 22.6831196 | 1.428163e-109 |
| DEW_POINT_TEMPERATURE | 632.384298 | 217.47791 | 2.9078093 | 3.652385e-03 |
| ‘19‘ | 519.455508 | 34.54060 | 15.0389836 | 3.001427e-50 |
| ‘8‘ | 502.353808 | 33.50621 | 14.9928559 | 5.875496e-50 |
| ‘21‘ | 430.185215 | 34.33684 | 12.5283880 | 1.380106e-35 |
| ‘20‘ | 394.634676 | 34.72886 | 11.3633072 | 1.234052e-29 |
| ‘4‘ | -389.558392 | 34.03286 | -11.4465363 | 4.834550e-30 |
| ‘5‘ | -379.055209 | 33.81771 | -11.2087790 | 6.910457e-29 |
| AUTUMN | 357.978265 | 19.99659 | 17.9019693 | 5.967468e-70 |
| ‘3‘ | -327.244955 | 33.88966 | -9.6561893 | 6.555260e-22 |
| ‘22‘ | 322.663949 | 33.93497 | 9.5083031 | 2.689464e-21 |
| ‘17‘ | 306.732435 | 34.27155 | 8.9500589 | 4.599800e-19 |
| ‘2‘ | -259.946174 | 34.57530 | -7.5182614 | 6.322812e-14 |
| SOLAR_RADIATION | 249.752454 | 41.34675 | 6.0404379 | 1.624462e-09 |
| SNOWFALL | 232.450790 | 104.60901 | 2.2220915 | 2.631243e-02 |
| ‘11‘ | -231.428411 | 33.98053 | -6.8106172 | 1.061643e-11 |
| ‘10‘ | -222.233262 | 33.31558 | -6.6705509 | 2.764896e-11 |
| ‘6‘ | -216.689105 | 34.16556 | -6.3423250 | 2.419940e-10 |
| (Intercept) | 216.584059 | 50.67322 | 4.2741329 | 1.947061e-05 |
| ‘12‘ | -194.605041 | 34.55857 | -5.6311662 | 1.867030e-08 |
| SPRING | 194.421437 | 19.21791 | 10.1166801 | 7.080230e-24 |
| ‘13‘ | -189.528426 | 35.19884 | -5.3845072 | 7.526455e-08 |
| ‘14‘ | -183.060291 | 35.24584 | -5.1938130 | 2.124823e-07 |
| SUMMER | 172.901302 | 29.04270 | 5.9533473 | 2.768259e-09 |
| HOLIDAY | -129.167078 | 22.44739 | -5.7542142 | 9.112956e-09 |
| ‘1‘ | -125.701499 | 34.57857 | -3.6352430 | 2.799291e-04 |
| ‘7‘ | 110.491959 | 33.93040 | 3.2564298 | 1.134185e-03 |
| ‘15‘ | -103.126885 | 35.09640 | -2.9383890 | 3.311170e-03 |
| ‘23‘ | 90.180803 | 33.97406 | 2.6544016 | 7.964847e-03 |
| ‘16‘ | 43.597895 | 34.23026 | 1.2736652 | 2.028290e-01 |
| VISIBILITY | 24.368049 | 20.23481 | 1.2042639 | 2.285328e-01 |
| ‘0‘ | -12.499804 | 34.01006 | -0.3675326 | 7.132341e-01 |
| WIND_SPEED | -9.313319 | 40.26779 | -0.2312846 | 8.171012e-01 |

[31]:
```
train_fit2 %>%
  tidy() %>%
  filter(!is.na(estimate)) %>%
  ggplot(aes(x = fct_reorder(term, abs(estimate)), y = abs(estimate))) +
  geom_bar(stat = "identity", fill = "black") +
  coord_flip() +
  theme(axis.text.y = element_text(angle = 10, colour = "black", size = 7)) +
  ylab("coef")
```

```
  xlab("recorde.variable.coef")
```

```
$x
[1] "recorde.variable.coef"

attr(,"class")
[1] "labels"
```




```
[ ]:
```