

Linear Regression

Data Set

h2o

```
ames <- AmesHousing::make_ames()
ames.h2o <- as.h2o(ames)
```

stratified (*Sale_Price*) training sample

```
set.seed(123)
```

```
split <- initial_split(ames, prop = 0.7,
                       strata = "Sale_Price")
```

```
ames_train <- training(split)
ames_test <- testing(split)
```

Simple Linear Model

```
model1 <- lm(Sale_Price ~ Gr_Liv_Area, data = ames_train)
```

```
# Fitted regression line (full training)
```

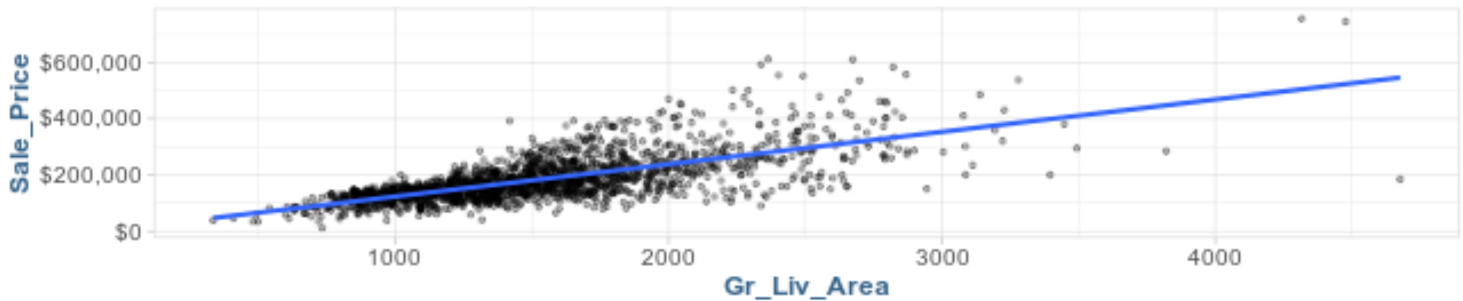
```
p1 <- model1 %>%
  broom::augment() %>%
  ggplot(aes(Gr_Liv_Area, Sale_Price)) +
  geom_point(size = 1, alpha = 0.3) +
  geom_smooth(se = F, method = "lm") +
  scale_y_continuous(labels = scales::dollar) +
  ggtitle("Fitted regression line")
```

```
# Fitted regression line (restricted range)
```

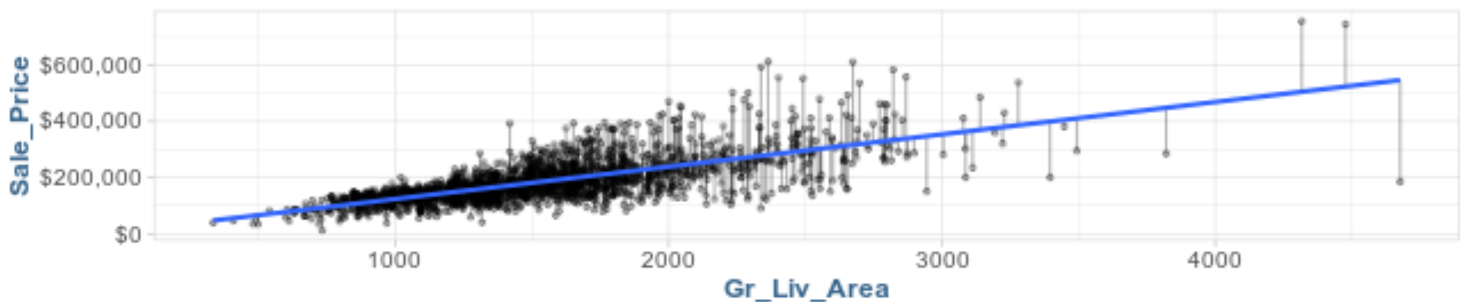
```
p2 <- model1 %>%
  broom::augment() %>%
  ggplot(aes(Gr_Liv_Area, Sale_Price)) +
  geom_segment(aes(x = Gr_Liv_Area, y = Sale_Price,
                  xend = Gr_Liv_Area, yend = .fitted),
              alpha = .3) +
  geom_point(size = 1, alpha = 0.3) +
  geom_smooth(se = F, method = "lm") +
  scale_y_continuous(labels = scales::dollar) +
  ggtitle("Fitted regression line (with residuals)")
```

```
grid.arrange(p1, p2, nrow = 2)
```

Fitted regression line



Fitted regression line (with residuals)



```
summary(model1)
```

Call:

```
lm(formula = Sale_Price ~ Gr_Liv_Area, data = ames_train)
```

Residuals:

Min	1Q	Median	3Q	Max
-361143	-30668	-2449	22838	331357

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	8732.938	3996.613	2.185	0.029 *
Gr_Liv_Area	114.876	2.531	45.385	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 56700 on 2051 degrees of freedom

Multiple R-squared: 0.5011, Adjusted R-squared: 0.5008

F-statistic: 2060 on 1 and 2051 DF, p-value: < 2.2e-16

```
[1] 56704.78
```

```
[1] 3215432370
```

Inference

The variability of an estimate is its *standard error (SE)*, the square root of its variance.

t-test for the coefficients are simply the estimated coefficient divided by the standard error (t value = Estimate / Std. Error)

t-test measure the number of standard deviations each coefficient is away from zero (basically $\text{abs}(T) > 2$ is significant at 95% conf)

The confidence interval for coefficients is:

$$\hat{\beta}_j \pm t_{1-\alpha/2, n-p} \hat{SE}(\hat{\beta}_j)$$

```
confint(model1, level = .95)
```

	2.5 %	97.5 %
(Intercept)	895.0961	16570.7805
Gr_Liv_Area	109.9121	119.8399