

# A tale of two variables

INTRODUCTION TO REGRESSION IN R



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# Swedish motor insurance data

- Each row represents one geographic region in Sweden.
- There are 63 rows.

n_claims	total_payment_sek
108	392.5
19	46.2
13	15.7
124	422.2
40	119.4
...	...

# Descriptive statistics

```
library(dplyr)
swedish_motor_insurance %>%
  summarize_all(mean)
```

```
# A tibble: 1 x 2
  n_claims total_payment_sek
  <dbl>      <dbl>
1    22.9      98.2
```

```
swedish_motor_insurance %>%
  summarize(
    correlation = cor(n_claims, total_payment_sek)
  )
```

```
# A tibble: 1 x 1
  correlation
  <dbl>
1    0.881
```

# What is regression?

- Statistical models to explore the relationship a response variable and some explanatory variables.
- Given values of explanatory variables, you can predict the values of the response variable.

n_claims	total_payment_sek
108	392.5
19	46.2
13	15.7
124	422.2
40	119.4
<b>200</b>	<b>???</b>

# Jargon

## **Response variable (a.k.a. dependent variable)**

The variable that you want to predict.

## **Explanatory variables (a.k.a. independent variables)**

The variables that explain how the response variable will change.

# Linear regression and logistic regression

## Linear regression

- The response variable is numeric.

## Logistic regression

- The response variable is logical.

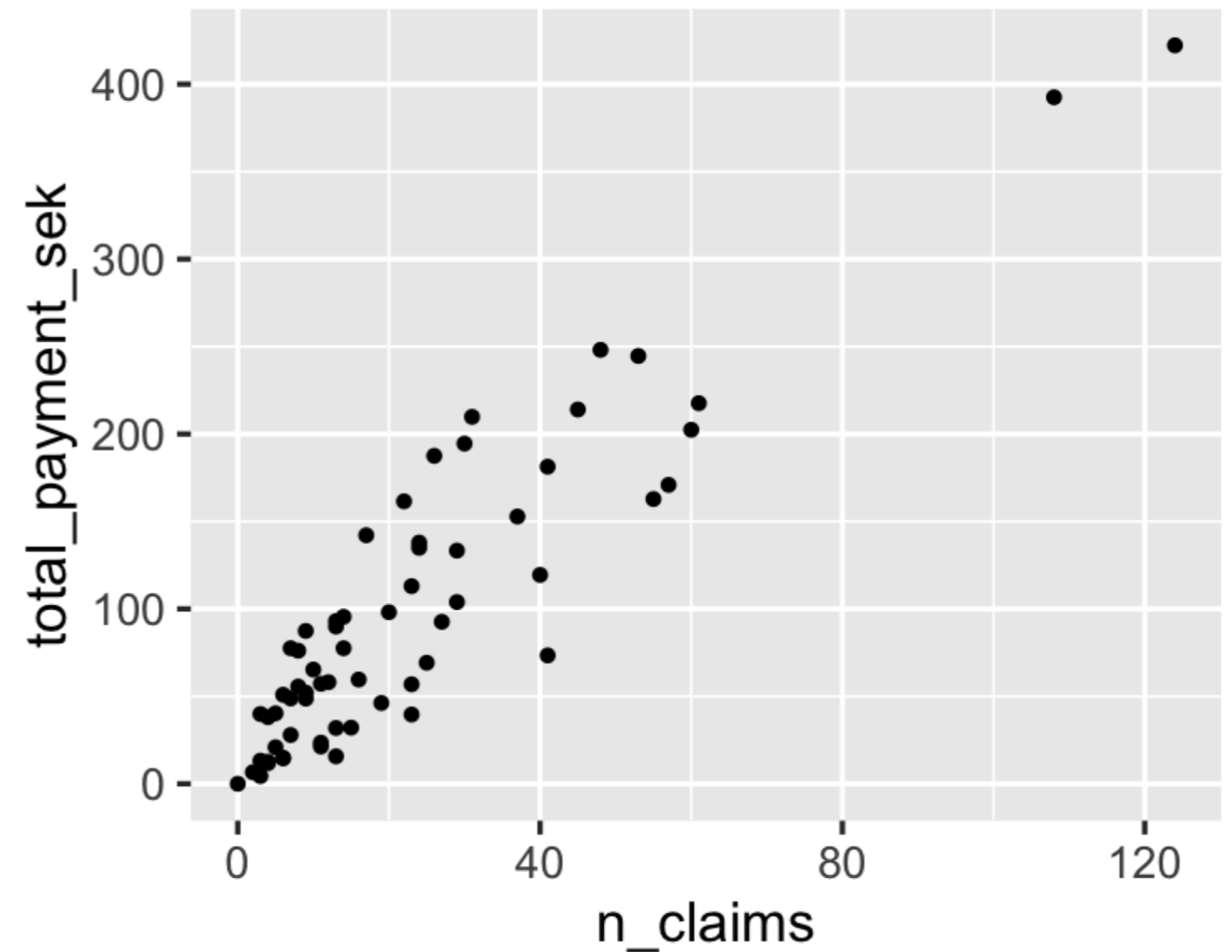
## Simple linear/logistic regression

- There is only one explanatory variable.

# Visualizing pairs of variables

```
library(ggplot2)

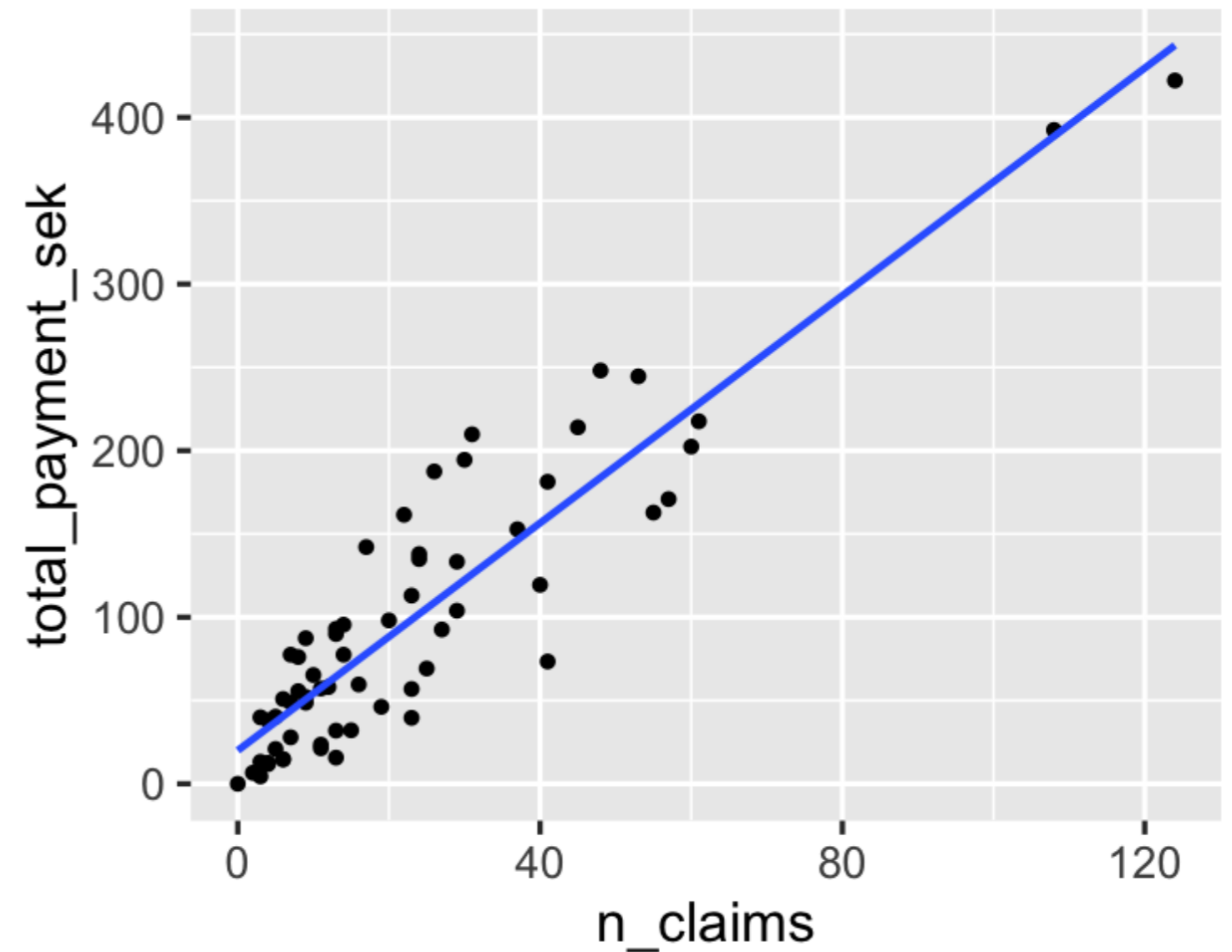
ggplot(
  swedish_motor_insurance,
  aes(n_claims, total_payment_sek)
) +
  geom_point()
```



# Adding a linear trend line

```
library(ggplot2)

ggplot(
  swedish_motor_insurance,
  aes(n_claims, total_payment_sek)
) +
  geom_point() +
  geom_smooth(
    method = "lm",
    se = FALSE
  )
```





# Course flow

## Chapter 1

Visualizing and fitting linear regression models.

## Chapter 2

Making predictions from linear regression models and understanding model coefficients.

## Chapter 3

Assessing the quality of the linear regression model.

## Chapter 4

Same again, but with logistic regression models

# Let's practice!

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# Fitting a linear regression

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# Straight lines are defined by two things

## Intercept

The  $y$  value at the point when  $x$  is zero.

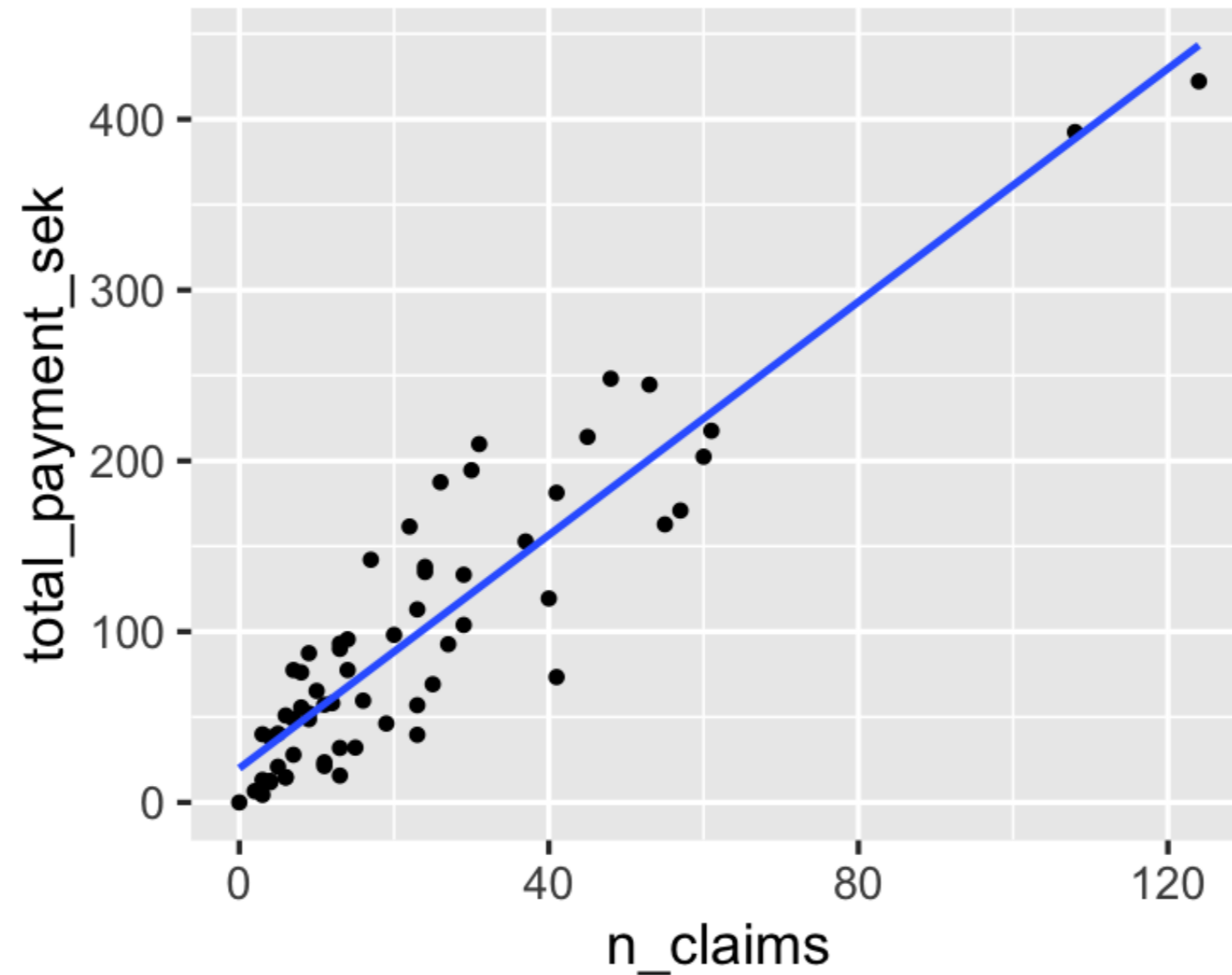
## Slope

The amount the  $y$  value increases if you increase  $x$  by one.

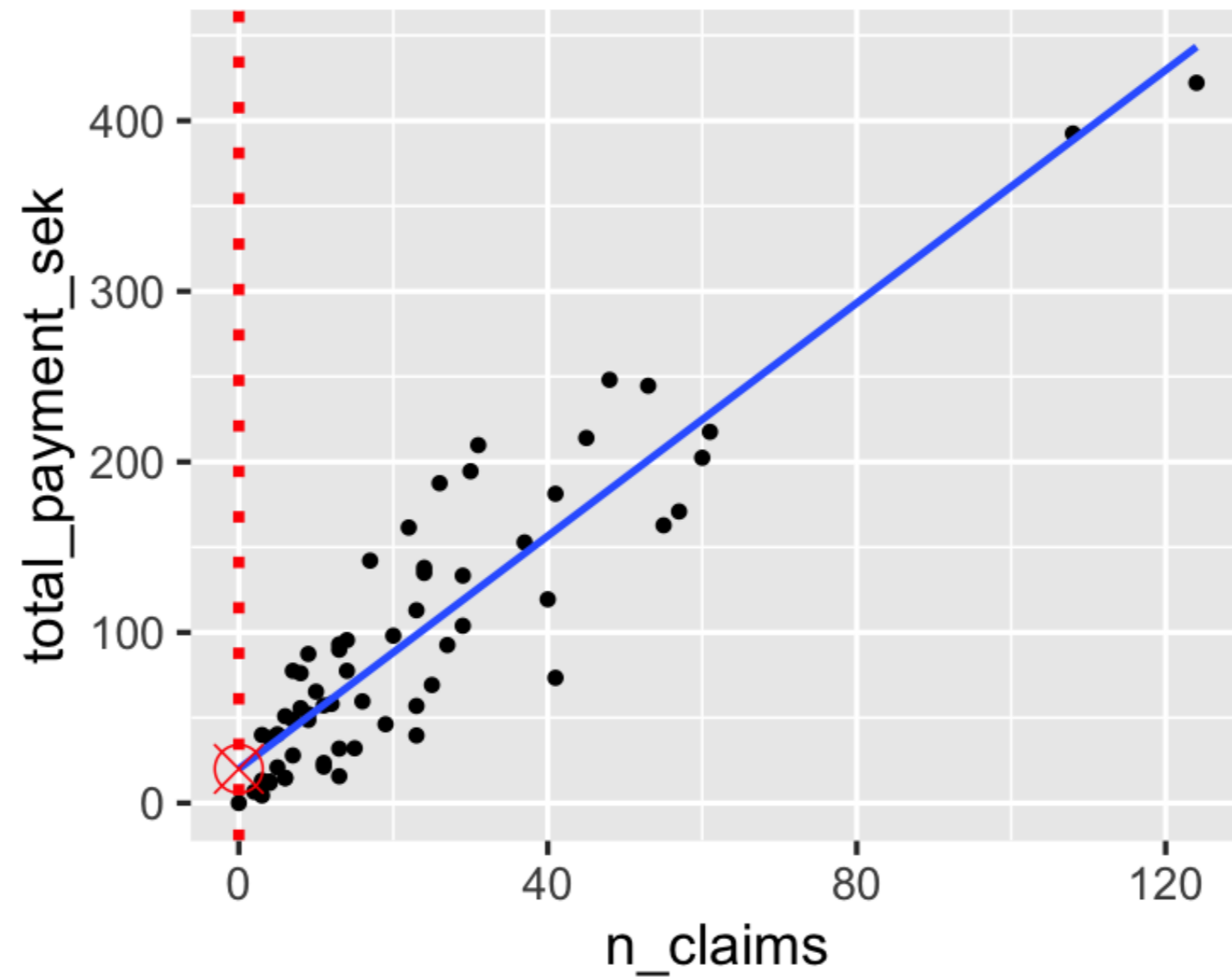
## Equation

$$y = \textit{intercept} + \textit{slope} * x$$

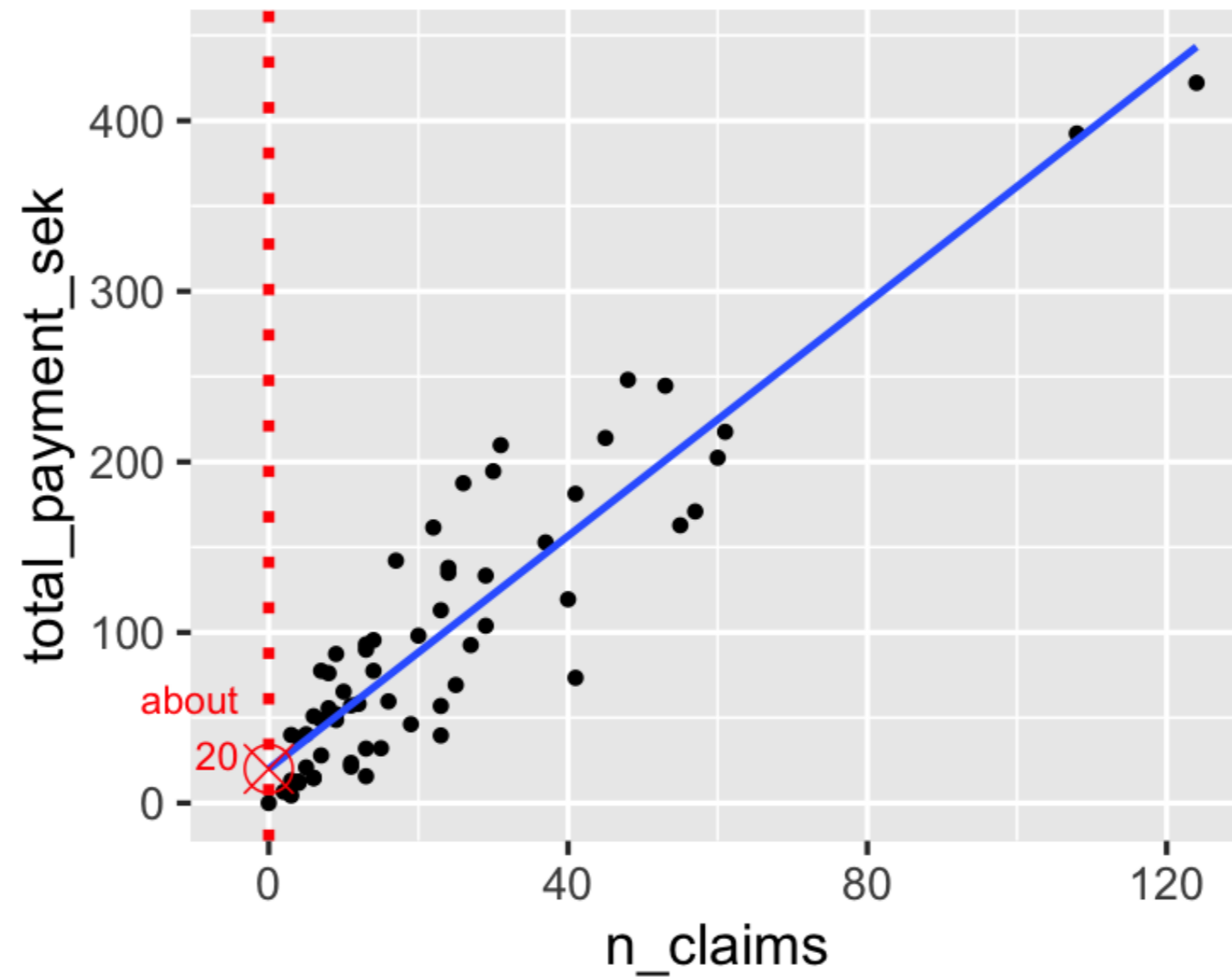
# Estimating the intercept



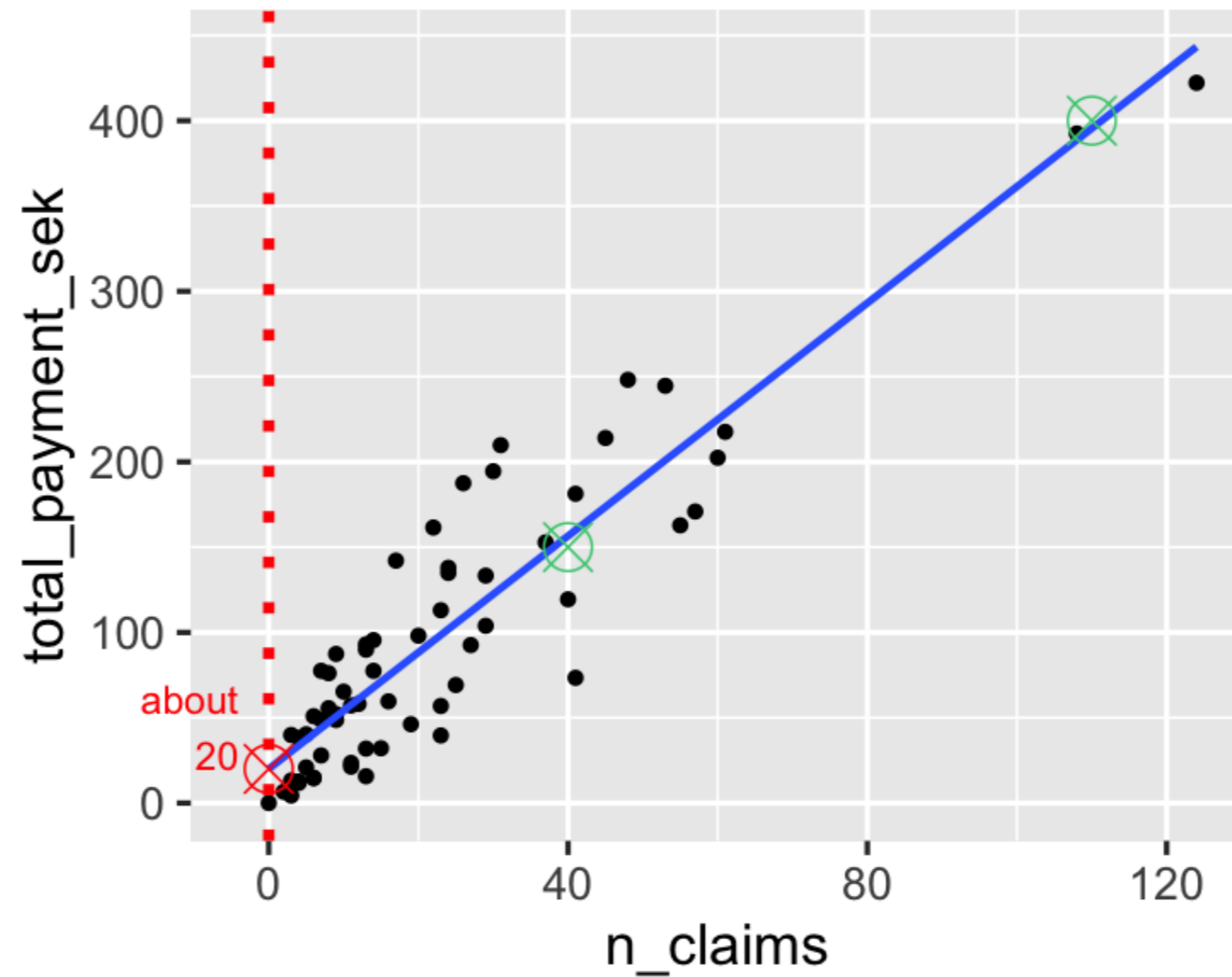
# Estimating the intercept



# Estimating the intercept

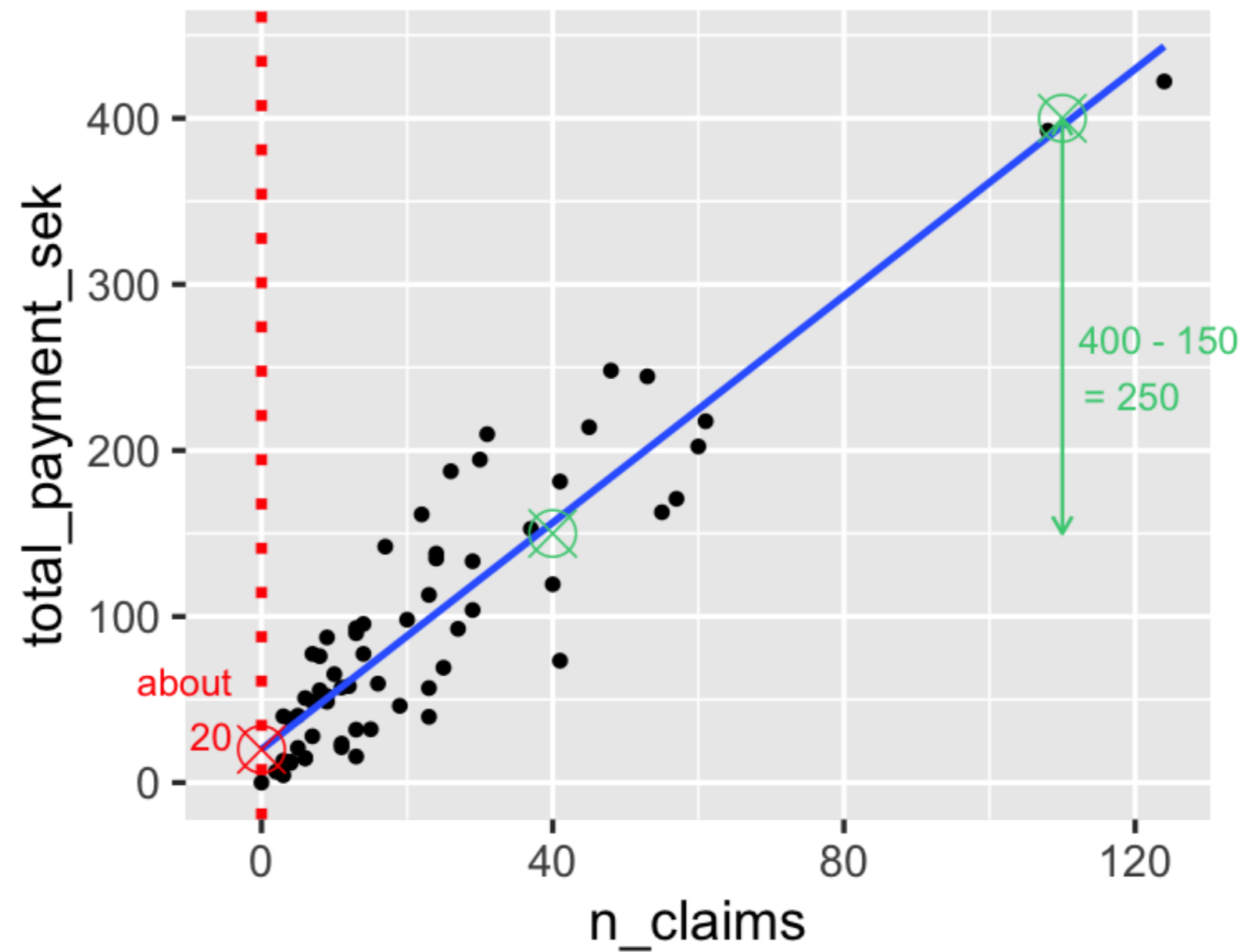


# Estimating the slope

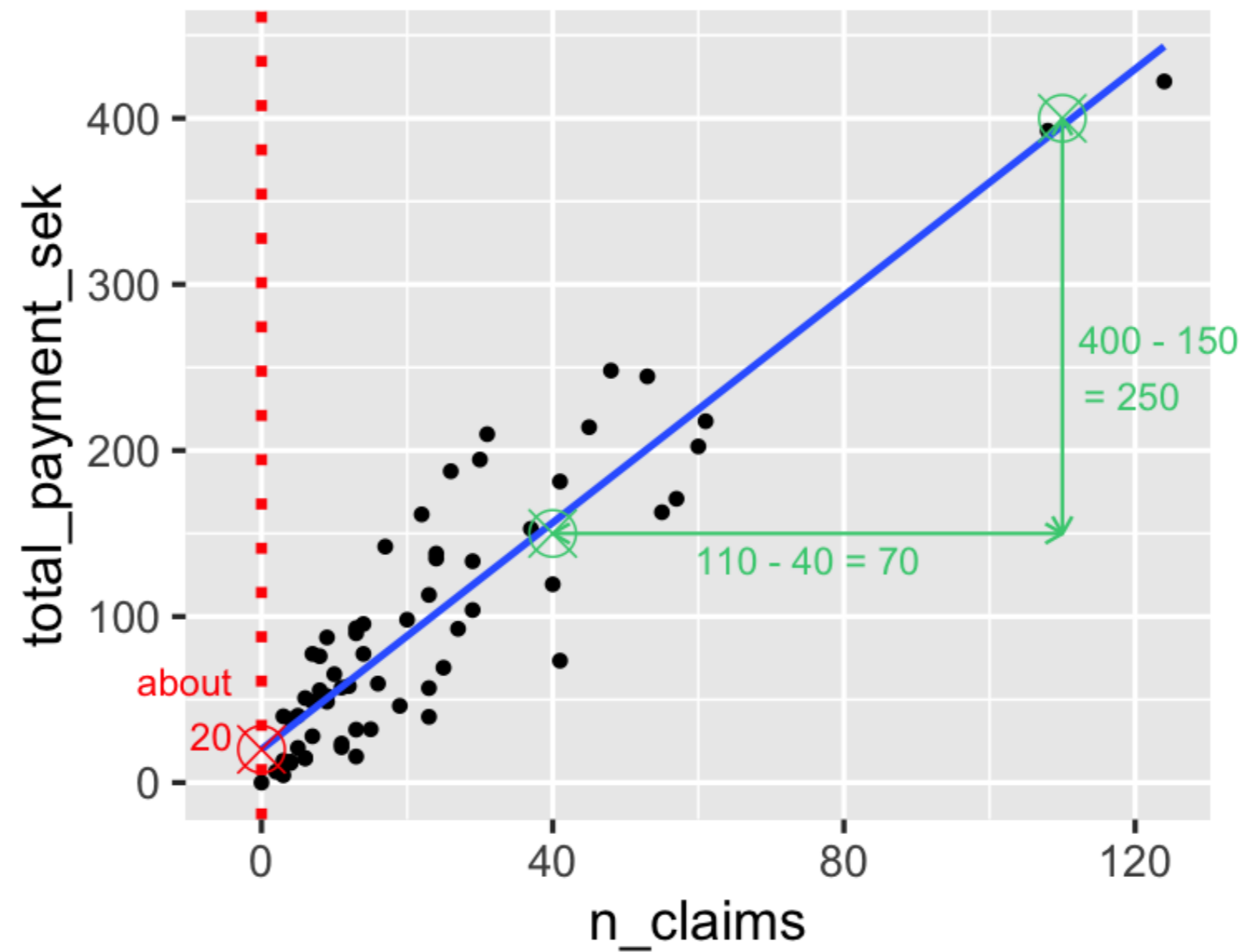




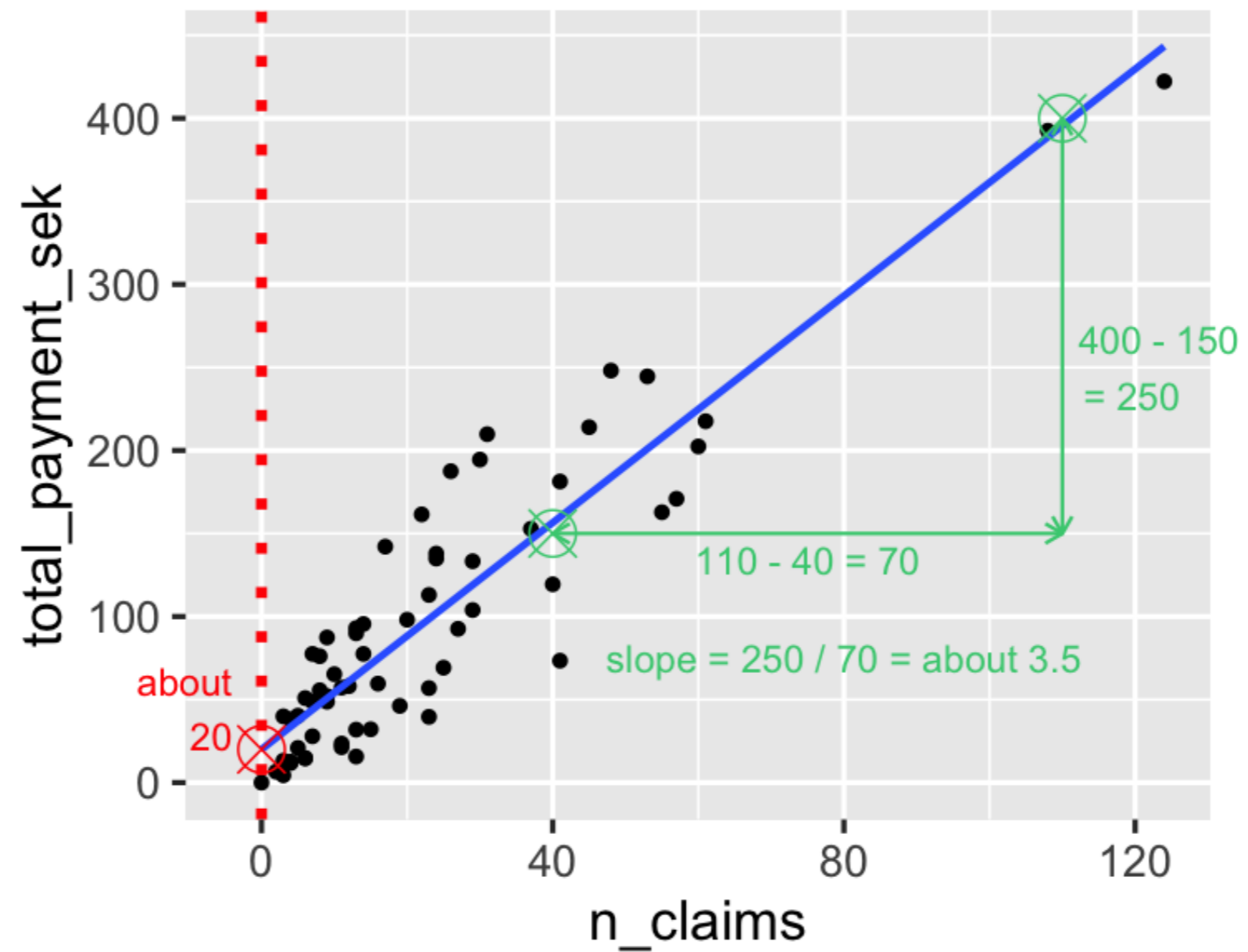
# Estimating the slope



# Estimating the slope



# Estimating the slope



# Running a model

```
lm(total_payment_sek ~ n_claims, data = swedish_motor_insurance)
```

Call:

```
lm(formula = total_payment_sek ~ n_claims, data = swedish_motor_insurance)
```

Coefficients:

(Intercept)	n_claims
19.994	3.414

# Interpreting the model coefficients

Call:

```
lm(formula = total_payment_sek ~ n_claims, data = swedish_motor_insurance)
```

Coefficients:

(Intercept)	n_claims
19.994	3.414

## Equation

$$total\_payment\_sek = 19.994 + 3.414 * n\_claims$$

# Let's practice!

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# Categorical explanatory variables

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# Fish dataset

- Each row represents one fish.
- There are 128 rows in the dataset.
- There are 4 species of fish.

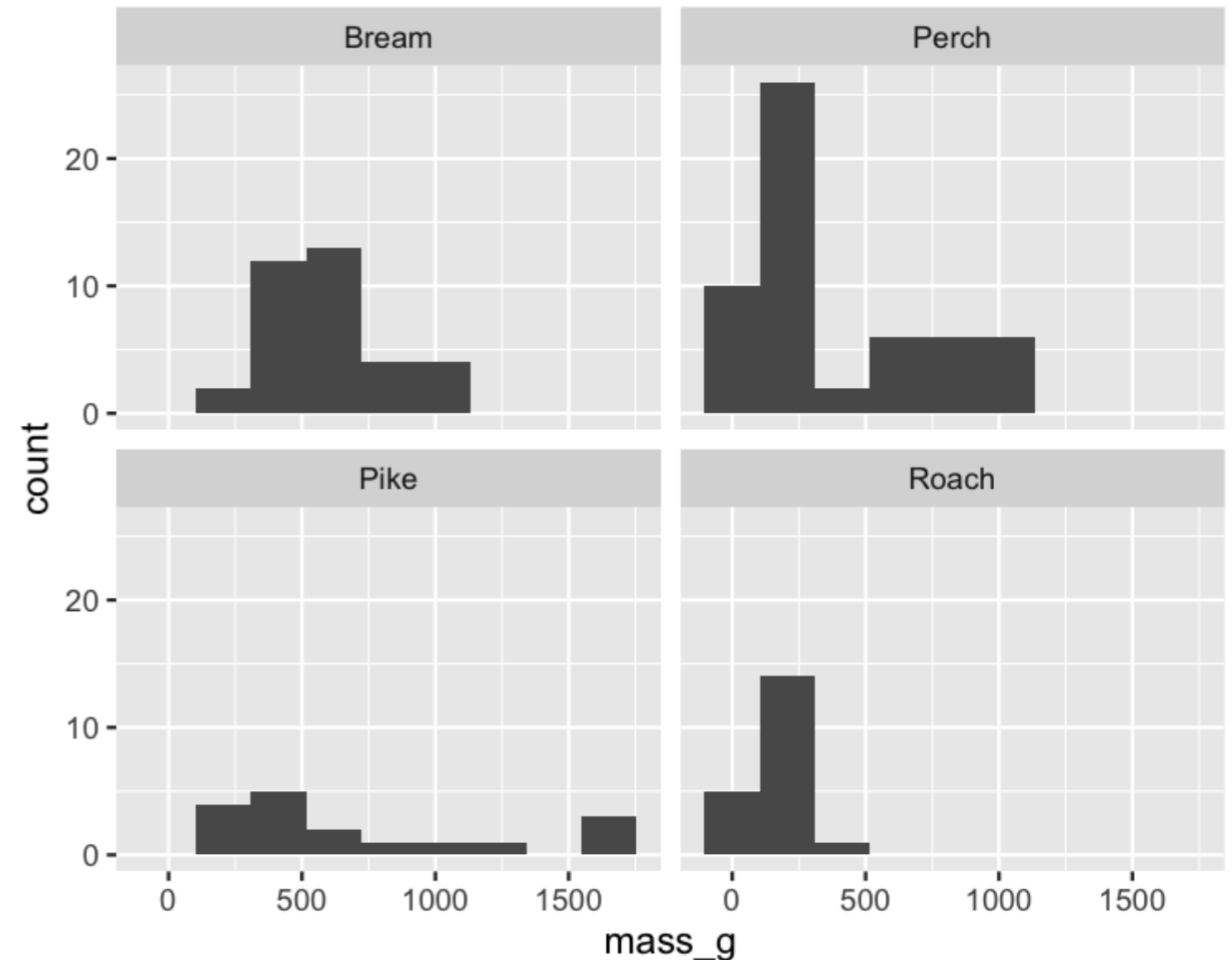
species	mass_g
Bream	242.0
Perch	5.9
Pike	200.0
Roach	40.0
...	...



# Visualizing 1 numeric and 1 categorical variable

```
library(ggplot2)
```

```
ggplot(fish, aes(mass_g)) +  
  geom_histogram(bins = 9) +  
  facet_wrap(vars(species))
```



# Summary statistics: mean mass by species

```
fish %>%  
  group_by(species) %>%  
  summarize(mean_mass_g = mean(mass_g))
```

```
# A tibble: 4 x 2  
  species mean_mass_g  
  <chr>      <dbl>  
1 Bream      618.  
2 Perch      382.  
3 Pike       719.  
4 Roach      152.
```

# Linear regression

```
lm(mass_g ~ species, data = fish)
```

Call:

```
lm(formula = mass_g ~ species, data = fish)
```

Coefficients:

(Intercept)	speciesPerch	speciesPike	speciesRoach
617.8	-235.6	100.9	-465.8

# No intercept

```
lm(mass_g ~ species + 0, data = fish)
```

Call:

```
lm(formula = mass_g ~ species + 0, data = fish)
```

Coefficients:

speciesBream	speciesPerch	speciesPike	speciesRoach
617.8	382.2	718.7	152.0

# Let's practice!

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