

Linear models 2

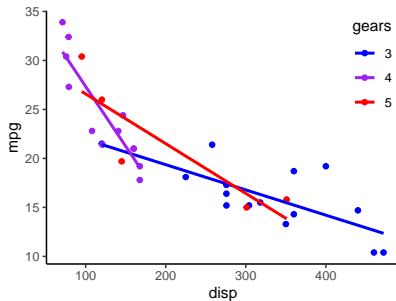
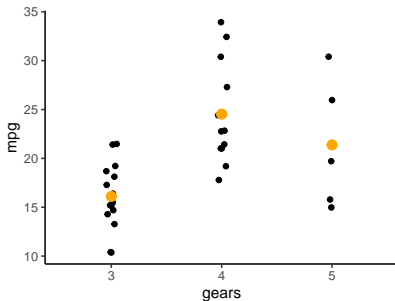
More bells and whistles

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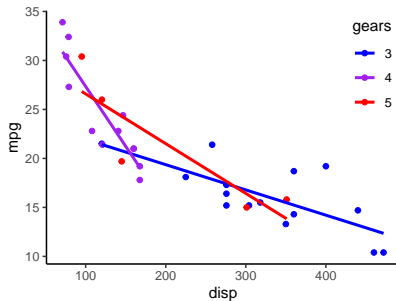
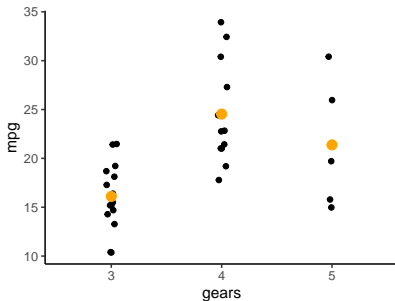
Motivation

- *I have 2+ groups of data, and I want to know whether the means are different*



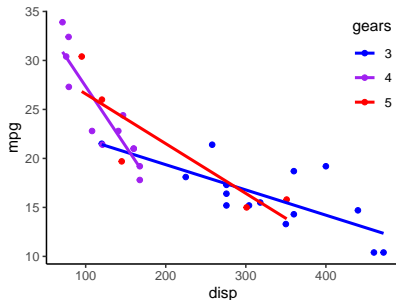
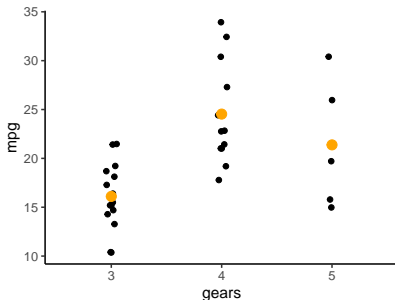
Motivation

- *I have 2+ groups of data, and I want to know whether the means are different*
- *I have 2+ groups of bivariate data, and I want to know whether the relationships differ between groups*

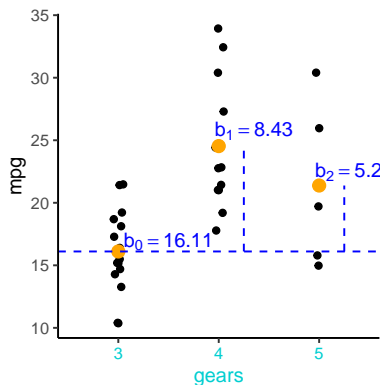


Motivation

- *I have 2+ groups of data, and I want to know whether the means are different*
- *I have 2+ groups of bivariate data, and I want to know whether the relationships differ between groups*
- **How do we know if any of this matters?**



Categorical data, 3 categories



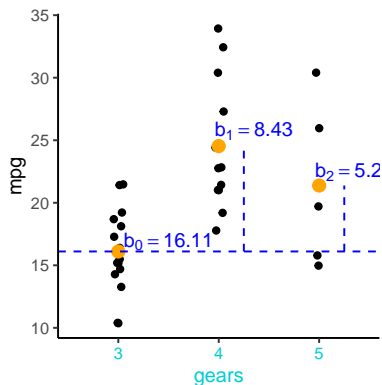
The more factor levels, the more coefficients:

- *mpg* is the thing you're interested in predicting

$$\hat{mpg} = b_0 + b_1 \text{gears}_4 + b_2 \text{gears}_5$$

$$mpg \sim \text{Normal}(\hat{mpg}, \sigma)$$

Categorical data, 3 categories



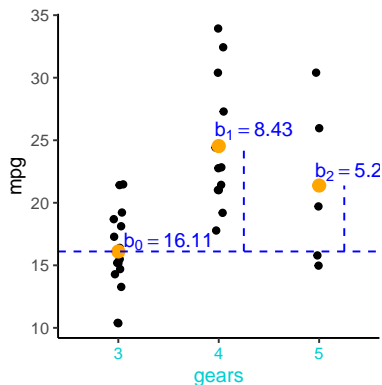
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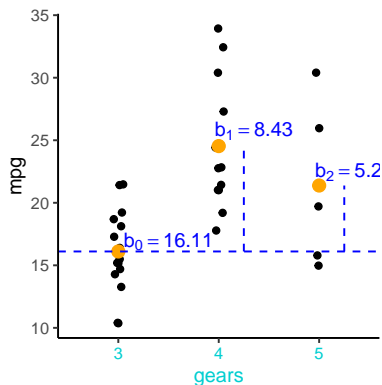
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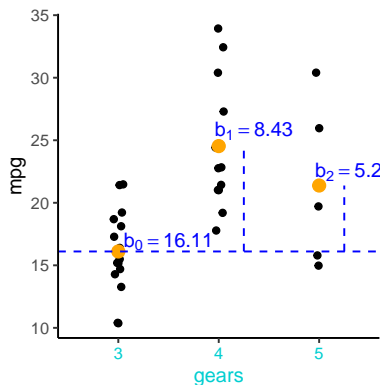
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- mpg is the thing you're interested in predicting
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- $gear$ is the *predictor* of mpg
- set of 0s and 1s

$$\hat{mpg} = b_0 + b_1 gear_4 + b_2 gear_5$$

$$mpg \sim Normal(\hat{mpg}, \sigma)$$

Categorical data, 3 categories



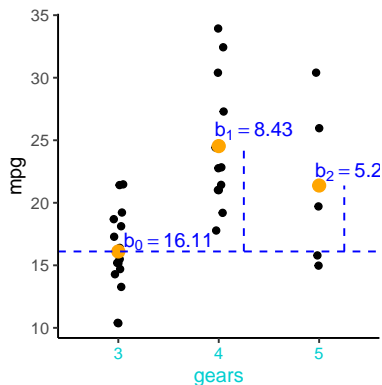
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- $gears_4$ = "is this data point from a 4-gear car?"

$$\hat{mpg} = b_0 + b_1 gears_4 + b_2 gears_5$$

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Categorical data, 3 categories



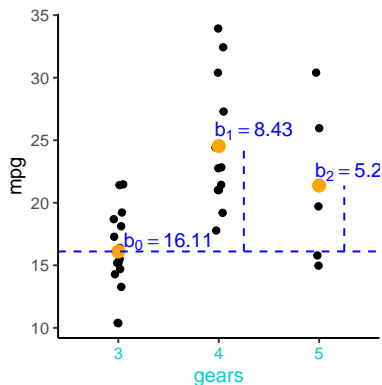
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- b_0 = *intercept*

$$\hat{mpg} = b_0 + b_1 gears_4 + b_2 gears_5$$

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Categorical data, 3 categories



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- \hat{mpg} is the *predicted value* of mpg
- $gear$ is the *predictor* of mpg
- set of 0s and 1s
- $gears_4$ = "is this data point from a 4-gear car?"
- b_0 = *intercept*
- $[b_1, b_2]$ = are *coefficients* for $gears$

$$\hat{mpg} = b_0 + b_1 gears_4 + b_2 gears_5$$

$$mpg \sim Normal(\hat{mpg}, \sigma)$$

How do I get R to fit this model?

```
#Formula structure: y ~ x
mod1 <- lm(mpg ~ factor(gear), #mpg depends on gears
           data = mtcars) #Name of the dataframe containing mpg & gears
summary(mod1)
```

```
##
## Call:
## lm(formula = mpg ~ factor(gear), data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.7333 -3.2333 -0.9067  2.8483  9.3667
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    16.107      1.216   13.250 7.87e-14 ***
## factor(gear)4     8.427      1.823    4.621 7.26e-05 ***
## factor(gear)5     5.273      2.431    2.169  0.0384 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.708 on 29 degrees of freedom
## Multiple R-squared:  0.4292, Adjusted R-squared:  0.3898
## F-statistic: 10.9 on 2 and 29 DF, p-value: 0.0002948
```

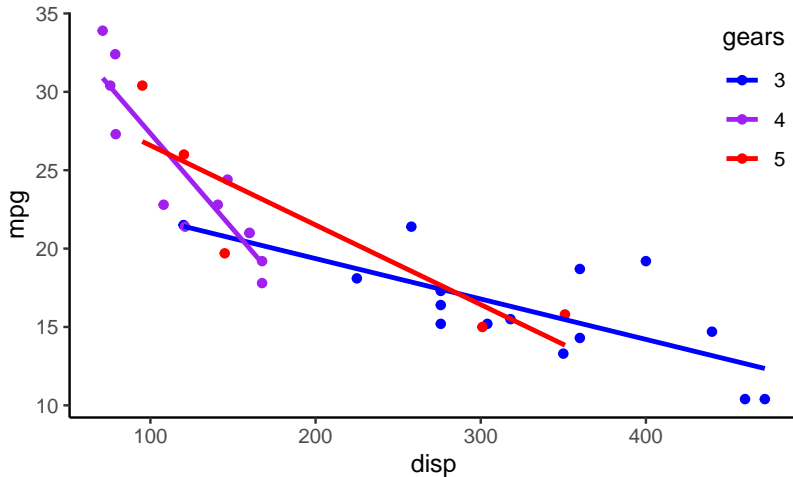
Dummy variables

```
mod1Matrix <- model.matrix(mod1) #Get model matrix (columns used to predict mpg)  
head(mod1Matrix,28) #Show first 28 rows of model matrix
```

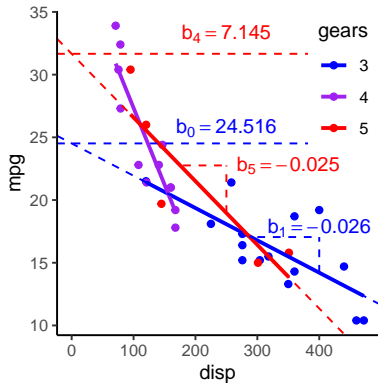
	(Intercept)	factor(gear)4	factor(gear)5
## Mazda RX4	1	1	0
## Mazda RX4 Wag	1	1	0
## Datsun 710	1	1	0
## Hornet 4 Drive	1	0	0
## Hornet Sportabout	1	0	0
## Valiant	1	0	0
## Duster 360	1	0	0
## Merc 240D	1	1	0
## Merc 230	1	1	0
## Merc 280	1	1	0
## Merc 280C	1	1	0
## Merc 450SE	1	0	0
## Merc 450SL	1	0	0
## Merc 450SLC	1	0	0
## Cadillac Fleetwood	1	0	0
## Lincoln Continental	1	0	0
## Chrysler Imperial	1	0	0
## Fiat 128	1	1	0
## Honda Civic	1	1	0
## Toyota Corolla	1	1	0
## Toyota Corona	1	0	0
## Dodge Challenger	1	0	0
## AMC Javelin	1	0	0
## Camaro Z28	1	0	0
## Pontiac Firebird	1	0	0
## Fiat X1-9	1	1	0
## Porsche 914-2	1	0	1
## Lotus Europa	1	0	1

Interactions

What if the slopes *and* intercepts differ between groups?



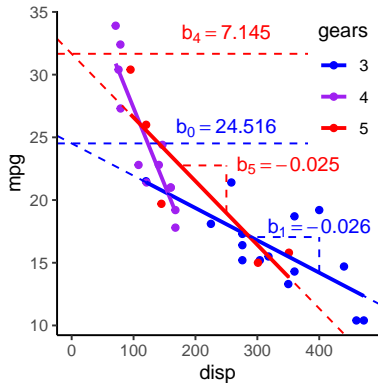
Interactions



$$\begin{aligned} \hat{mpg} = & b_0 + b_1 disp \\ & + b_2 gears_4 + b_3 gears_5 \\ & + b_4 (disp \times gears_4) \\ & + b_5 (disp \times gears_5) \\ mpg \sim & Normal(\hat{mpg}, \sigma) \end{aligned}$$

- Interactions occur when predictors are *multiplied*

Interactions



$$\begin{aligned} \hat{mpg} &= b_0 + b_1 disp \\ &+ b_2 gears_4 + b_3 gears_5 \\ &+ b_4 (disp \times gears_4) \\ &+ b_5 (disp \times gears_5) \\ mpg &\sim Normal(\hat{mpg}, \sigma) \end{aligned}$$

- Interactions occur when predictors are *multiplied*
- In this case, *disp* is multiplied by *gears₄* and *gears₅*

How do I get R to fit this model?

```
#Formula structure: y ~ x
mod2 <- lm(mpg ~ disp*factor(gear), #mpg depends on disp interacted with gears
           data = mtcars) #Name of the dataframe
summary(mod2)
```

```
##
## Call:
## lm(formula = mpg ~ disp * factor(gear), data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.5986 -1.5990 -0.0143  1.6329  4.9926
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    24.51556     2.462431   9.956 2.32e-10 ***
## disp          -0.025770    0.007265  -3.547 0.001505 **
## factor(gear)4    15.051963    3.558043   4.230 0.000256 ***
## factor(gear)5     7.145380    3.535913   2.021 0.053711 .
## disp:factor(gear)4 -0.096442    0.021261  -4.536 0.000114 ***
## disp:factor(gear)5 -0.025005    0.013320  -1.877 0.071742 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.579 on 26 degrees of freedom
## Multiple R-squared:  0.8465, Adjusted R-squared:  0.817
## F-statistic: 28.67 on 5 and 26 DF, p-value: 8.452e-10
```

Beware of fitting too many interactions, or else the *Bilbo effect* occurs!

Dummy variables

```
mod2Matrix <- model.matrix(mod2) #Get model matrix (columns used to predict mpg)  
colnames(mod2Matrix) <- gsub('factor\\((gear\\)', 'gear', colnames(mod2Matrix)) #Shorten colnames  
head(mod2Matrix, 28) #Show first 28 rows of model matrix
```

```
##                (Intercept)  disp gear4 gear5 disp:gear4 disp:gear5  
## Mazda RX4                1 160.0      1      0      160.0      0.0  
## Mazda RX4 Wag            1 160.0      1      0      160.0      0.0  
## Datsun 710                1 108.0      1      0      108.0      0.0  
## Hornet 4 Drive            1 258.0      0      0        0.0      0.0  
## Hornet Sportabout        1 360.0      0      0        0.0      0.0  
## Valiant                   1 225.0      0      0        0.0      0.0  
## Duster 360                1 360.0      0      0        0.0      0.0  
## Merc 240D                 1 146.7      1      0      146.7      0.0  
## Merc 230                   1 140.8      1      0      140.8      0.0  
## Merc 280                   1 167.6      1      0      167.6      0.0  
## Merc 280C                  1 167.6      1      0      167.6      0.0  
## Merc 450SE                 1 275.8      0      0        0.0      0.0  
## Merc 450SL                 1 275.8      0      0        0.0      0.0  
## Merc 450SLC                1 275.8      0      0        0.0      0.0  
## Cadillac Fleetwood        1 472.0      0      0        0.0      0.0  
## Lincoln Continental        1 460.0      0      0        0.0      0.0  
## Chrysler Imperial         1 440.0      0      0        0.0      0.0  
## Fiat 128                   1  78.7      1      0       78.7      0.0  
## Honda Civic                1  75.7      1      0       75.7      0.0  
## Toyota Corolla             1  71.1      1      0       71.1      0.0  
## Toyota Corona              1 120.1      0      0        0.0      0.0  
## Dodge Challenger           1 318.0      0      0        0.0      0.0  
## AMC Javelin                1 304.0      0      0        0.0      0.0  
## Camaro Z28                 1 350.0      0      0        0.0      0.0  
## Pontiac Firebird           1 400.0      0      0        0.0      0.0  
## Fiat X1-9                   1  79.0      1      0       79.0      0.0  
## Porsche 914-2              1 120.3      0      1        0.0     120.3  
## Lotus Europa               1  95.1      0      1        0.0     95.1
```

How do I plot these model results?

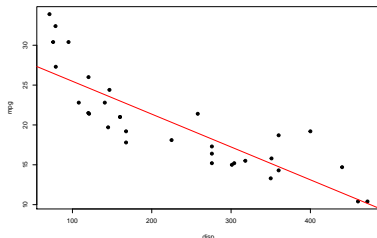
- If you have *1 variable* that you are using in your model, then boxplots and biplots of raw data are OK
- e.g. $mpg \sim disp$ or $mpg \sim gear$
- If you have *2+ variables* in your model, then avoid plotting raw data, and consider a **partial effects plot** instead
- e.g. $mpg \sim disp + gear$ or $mpg \sim disp * gear$

This model only has 1 variable, so plots of raw data are fine:

```
#Fit a model with 1 variable
mod3 <- lm(mpg~disp,data=mtcars)

#Plot raw data
plot(mpg ~ disp, data=mtcars,pch=19)

#Plot model fit (single line)
abline(mod3,col='red')
```



Problems with plotting raw data

- Say that I've fit the following model: `mpg ~ disp * gear`
- All of the plots below are using raw data, but which one is "telling the truth"?