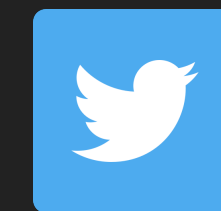


LINEAR MODELS ARE GREAT

ALEX GOLD
SOLUTIONS ENGINEER
RSTUDIO



@alexkgold

Slides at: https://github.com/akgold/dsdc_linear_models

IT ME.



WHAT DOES IT MEAN TO BE LINEAR?

21.0
21.0
22.8
21.4
18.7
18.1
14.3
24.4
22.8
19.2
17.8
16.4
17.3
15.2
10.4
10.4
14.7
32.4
30.4
33.9
21.5
15.5
15.2
13.3
19.2
27.3
26.0
30.4
15.8
19.7
15.0
21.4

$$= \beta_0 + \beta_{cyl}$$



Just a number

6
6
4
6
8
6
8
4
4
6
6
8
8
8
8
8
8
8
4
4
4
4
8
6
8
4

$$+ \beta_{hp}$$

160.0
160.0
108.0
258.0
360.0
225.0
360.0
146.7
140.8
167.6
167.6
275.8
275.8
275.8
472.0
460.0
440.0
78.7
75.7
71.1
120.1
318.0
304.0
350.0
400.0
79.0
120.3
95.1
351.0
145.0
301.0
121.0

+ some prediction error

WHAT DOES IT MEAN TO BE LINEAR?

$$Y = \beta X + \epsilon$$

ALEX <3 LINEAR
MODELS

```
> lm(mpg ~ cyl + disp + hp, data = mtcars)
```

Call:


```
lm(formula = mpg ~ cyl + disp + hp, data = mtcars)
```

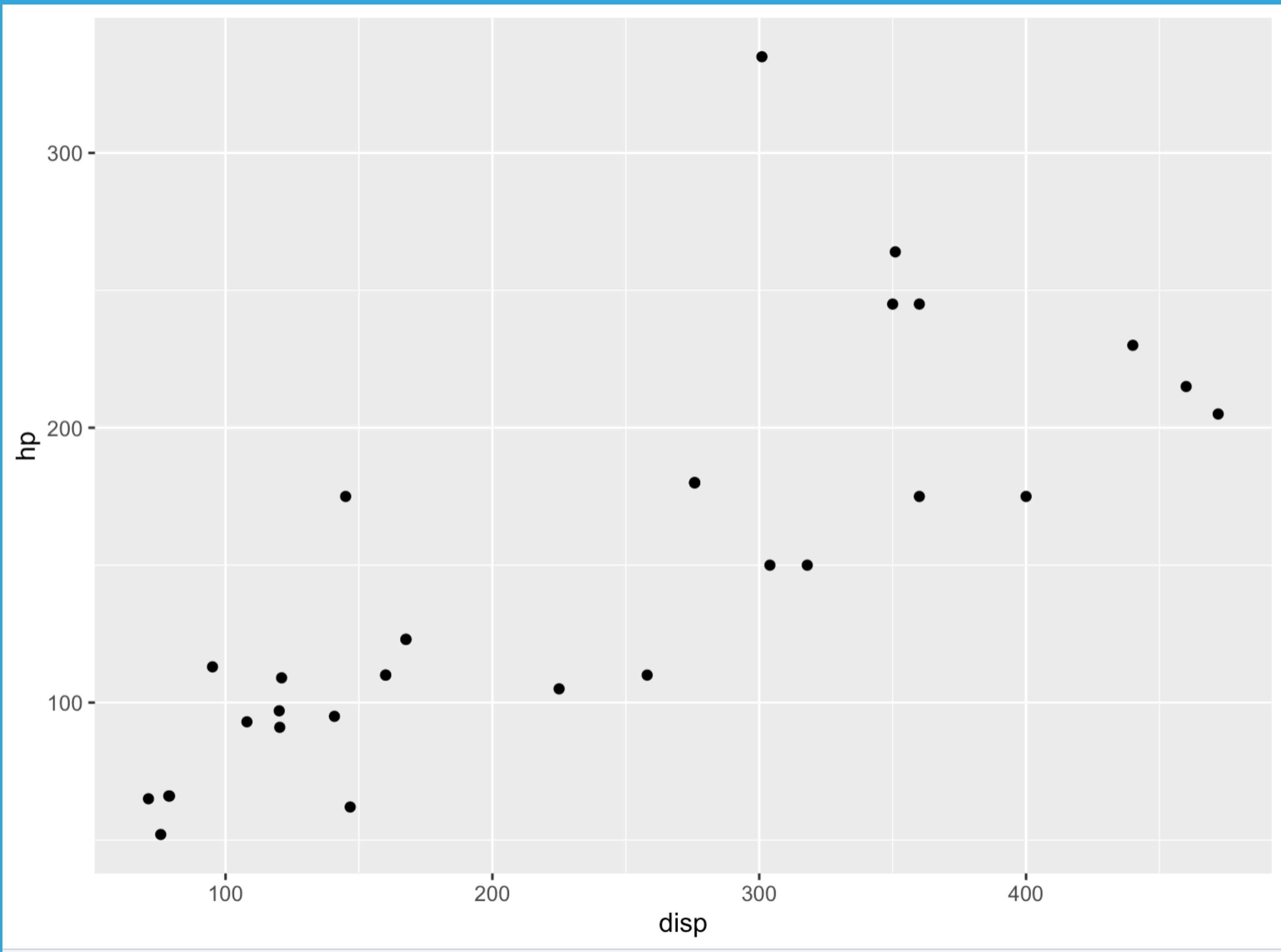
Coefficients:

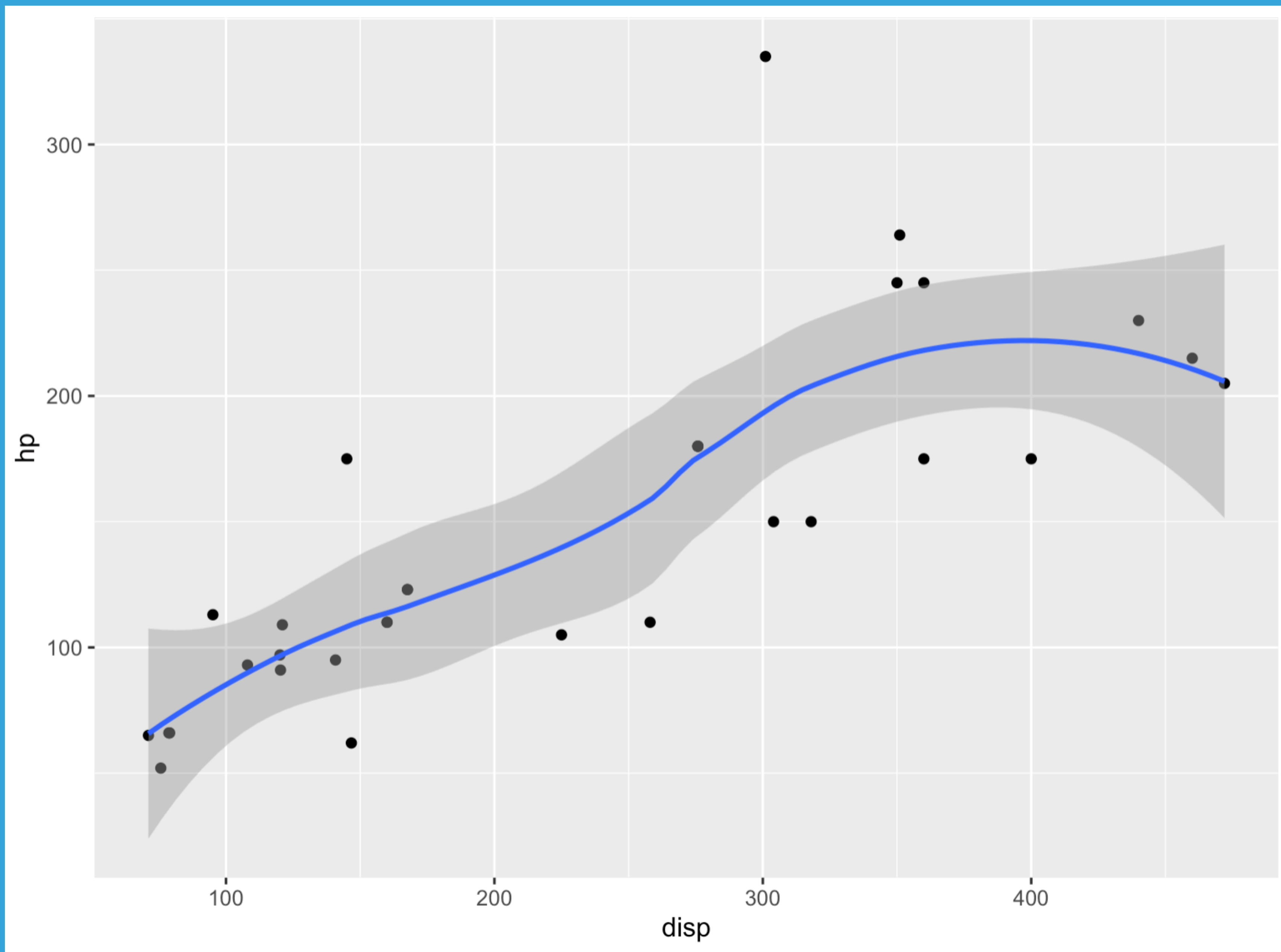
(Intercept)	cyl	disp	hp
34.18492	-1.22742	-0.01884	-0.01468

$$Y = \beta X + \epsilon$$

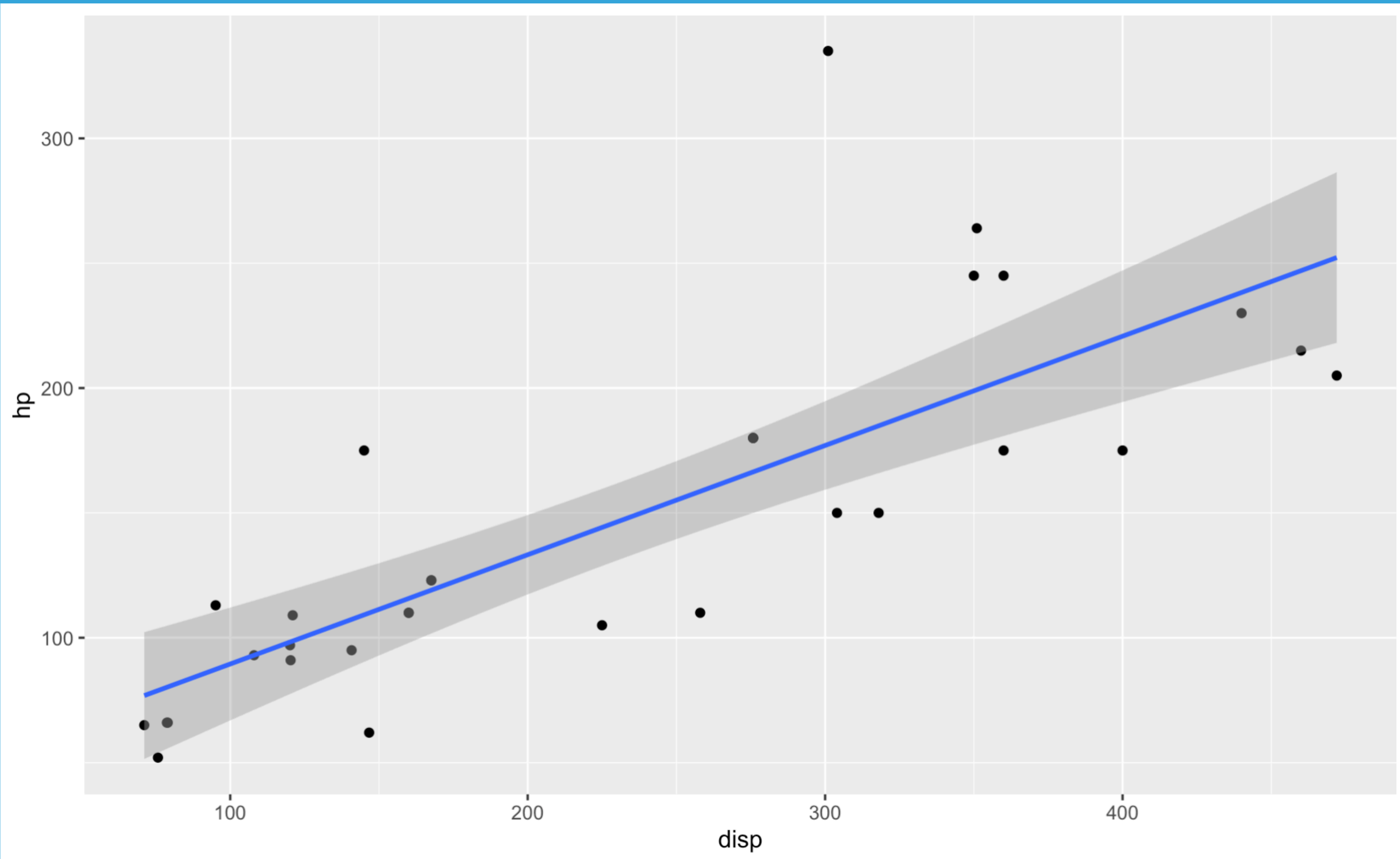


$$Y = \sum_k \beta_k f_k(x_i) + \epsilon$$
A red arrow points from the upper right towards the function f_k in the summation term of the equation.





```
ggplot(mtcars, aes(x = disp, y = hp)) + geom_point() + stat_smooth(method = "loess")
```

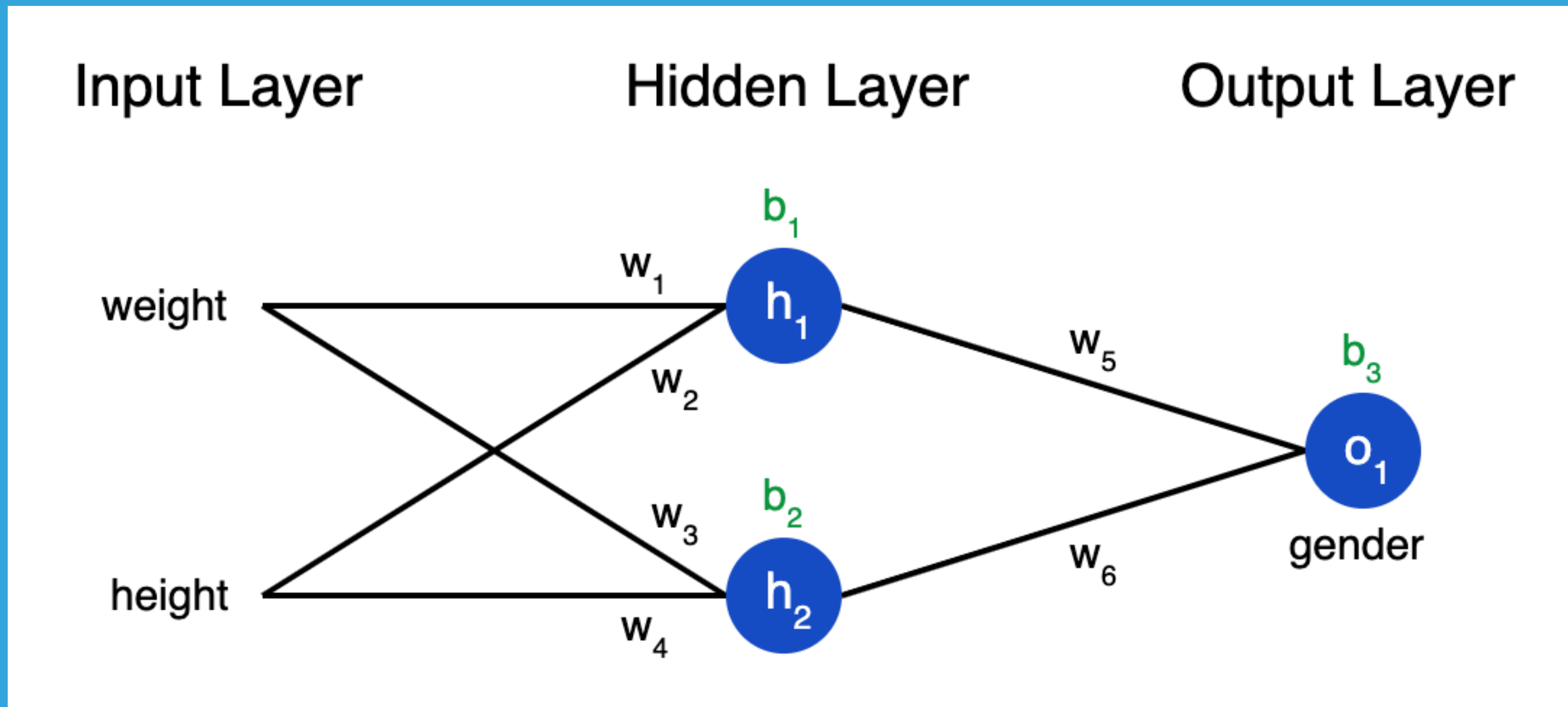


```
ggplot(mtcars, aes(x = disp, y = hp)) + geom_point() + stat_smooth(method = "lm")
```

$$\textit{gender} = \beta_0 + \beta_1 \textit{weight} + \beta_2 \textit{height}$$

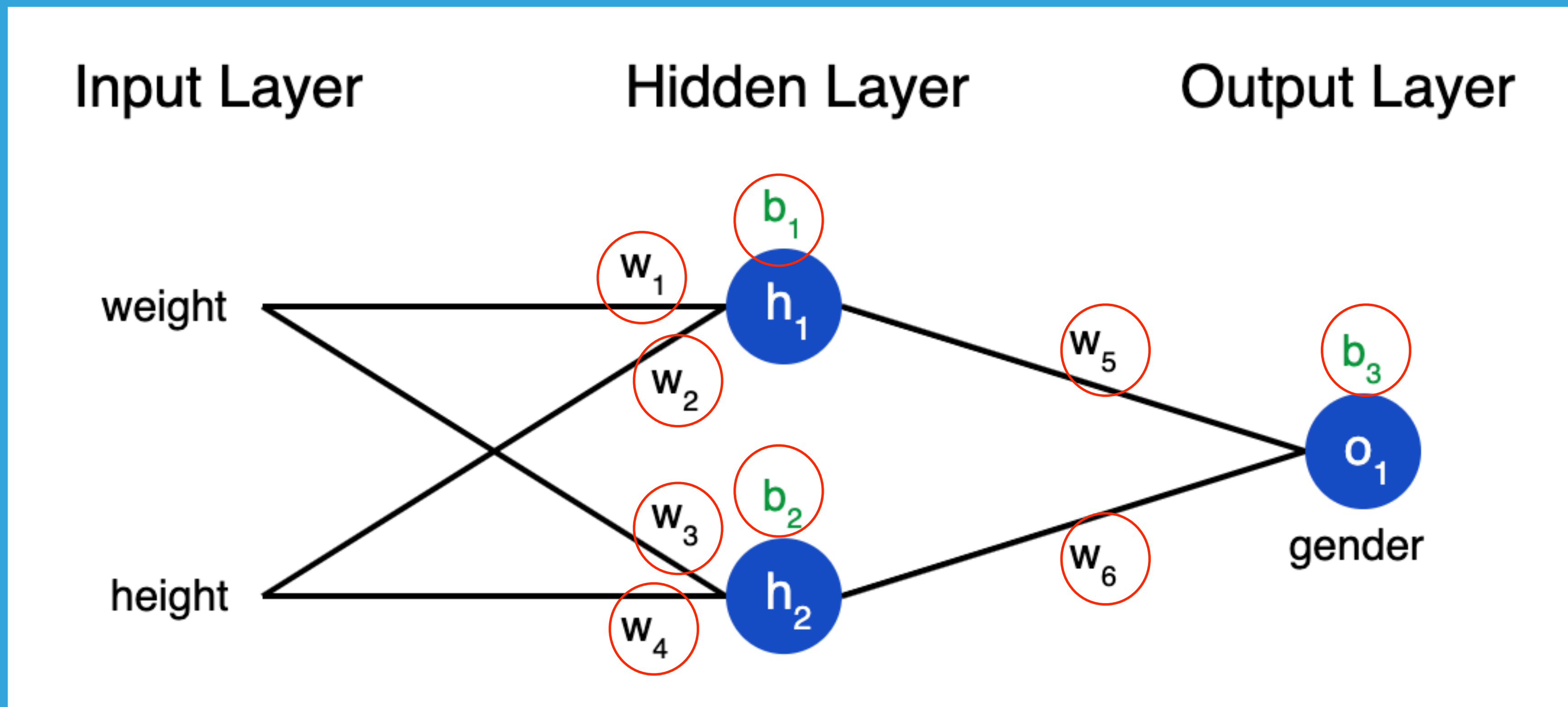
$$gender = \beta_0 + \beta_1 weight + \beta_2 height$$

VS



$$gender = \beta_0 + \beta_1 weight + \beta_2 height$$

VS





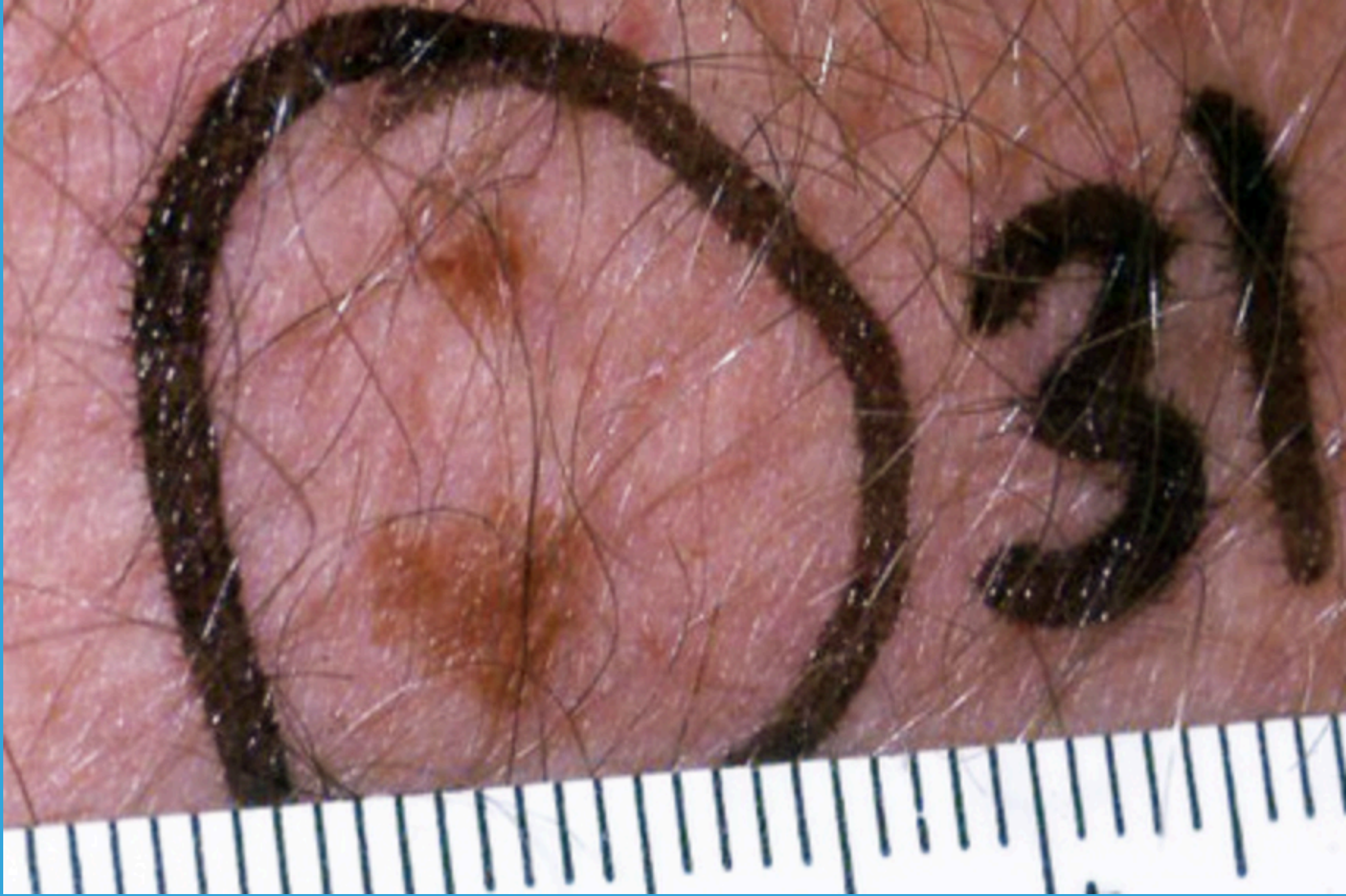
It's all about the Data-Generating Process

$$mpg = \beta_0 + \beta_1 cyl + disp$$

OR

$$mpg = \beta_0 + \beta_1 cyl + \beta_2 cyl^2 + \beta_3 \log(disp)$$

?



1. INTERPRETATION MATTERS.
2. LINEARITY ISN'T RESTRICTIVE.
3. MO' SQUIGGLY = MO' OVERFITTING.
4. SMALL DATA'S OK.
5. IT'S ALL ABOUT THE DGP.



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