



*NOTE:* From 2:22 onward, the slide title should say "Mean Absolute Error".

## Development of the derivative of the error function

Notice that we've defined the squared error to be

$$Error = \frac{1}{2}(y - \hat{y})^2.$$

Also, we've defined the prediction to be

$$\hat{y} = w_1 x + w_2.$$

So to calculate the derivative of the Error with respect to  $w_1$ , we simply use the chain rule:

$$\frac{\partial}{\partial w_1} Error = \frac{\partial Error}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial w_1}.$$

The first factor of the right hand side is the derivative of the Error with respect to the prediction  $\hat{y}$ , which is  $-(y - \hat{y})$ .

The second factor is the derivative of the prediction with respect to  $w_1$ , which is simply  $x$ .



$$\frac{\partial}{\partial w_1} \text{Error} = -(y - \hat{y})x$$

## Exercise

Calculate the derivative of the Error with respect to  $w_2$  and verify that it is precisely  $-(y - \hat{y})$ .

[NEXT](#)