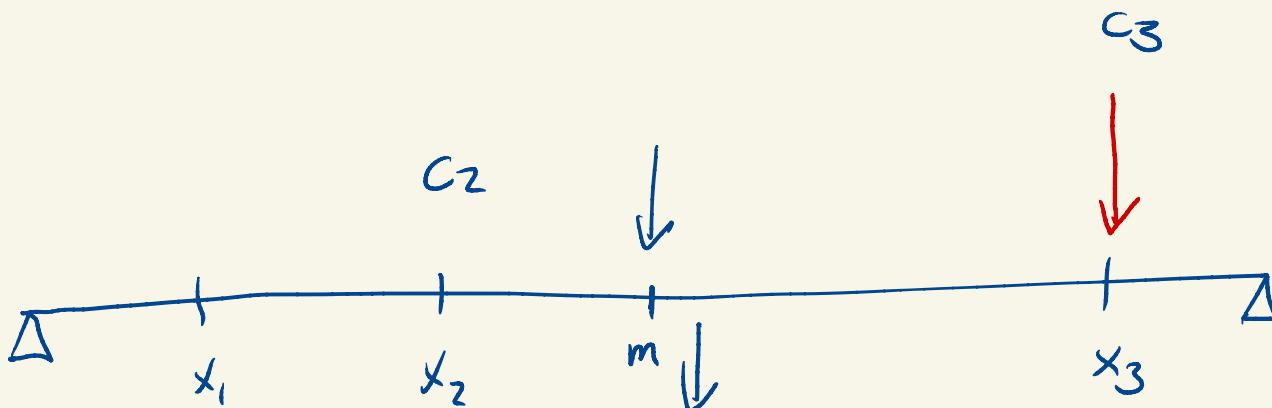


$c_1, c_2, c_3$



Want to measure the sag of the beam at  $m$

as a response to loads at  $x_1, x_2, x_3$

Say  $s$  is measured in mm.

Forces will be encoded as masses. (metric tonnes)

Masses:  $m_1, m_2, m_3$

$$s_{\text{say}} = \underbrace{c_1 m_1 + c_2 m_2 + c_3 m_3}_{s(m_1, m_2, m_3)} \quad \text{MS is a linear function.}$$
$$c^T m$$
$$c = (c_1, c_2, c_3)$$

where the coefficients  $c_i$  are known as  $m = (m_1, m_2, m_3)$   
sensitivities and have units of mm/tonne

$$f(x+y) = f(x) + f(y)$$

$$f(\alpha x) = \alpha f(x)$$

$$f(\alpha x + \beta y) = \alpha f(x) + \beta f(y)$$

$\alpha, \beta \in \mathbb{R}$

"superposition"

Claim: every linear function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  can be written in the form "f is a function from  $\mathbb{R}^n$  to  $\mathbb{R}$ "

$$f(x) = c^T x \quad \text{for some fixed vector } c,$$

$$e_k = (0, \dots, 0, 1, 0, \dots, 0)$$

$f \leftarrow$  we know it is linear.

$$c_k = f(e_k) \quad k=1, \dots, n$$

↗

(all these vectors  
have length  $n$ )

$$x = (x_1, x_2, \dots, x_n)$$

$$= x_1 e_1 + x_2 e_2 + \dots + x_n e_n$$

$$\left( = x_1 (1, 0, \dots, 0) + x_2 (0, 1, 0, \dots, 0) + \dots + x_n (0, 0, \dots, 0, 1) \right)$$

$$\left( = (x_1, 0, \dots, 0) + (0, x_2, 0, \dots, 0) + \dots + (0, 0, \dots, 0, x_n) \right)$$

$$\left( = (x_1, x_2, \dots, x_n) \right)$$

$$= x$$

$$x = (x_1, x_2, \dots, x_n)$$

$$(e_3)_4 = 0$$

$$= x_1 e_1 + x_2 e_2 + \dots + x_n e_n$$

$$f(x) = f(x_1 e_1 + x_2 e_2 + \dots + x_n e_n)$$

$$= f(x_1 e_1) + f(x_2 e_2) + \dots + f(x_n e_n)$$

$$= x_1 \boxed{f(e_1)} + x_2 f(e_2) + \dots + x_n f(e_n)$$

$$= x_1 c_1 + x_2 c_2 + \dots + x_n c_n$$

$$= c^T x$$

$$y = mx + b$$

$$f(x) = 3x - 7$$

$$y = 3x - 7$$

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

$$f(x+w) = f(x) + f(w)$$

$$f(\alpha x) = \alpha f(x)$$

$$\begin{aligned} f(1) &= -4 \\ f(0) &= -7 \\ f(0+1) &= f(0) + f(1) \\ f(1) &= -11 \\ -4 & \end{aligned}$$

$f \leftarrow$  linear

$$\begin{aligned}f(0) &= f(0+0) \\&= f(0) + f(0)\end{aligned}$$

$$f(0) = 2 f(0)$$

$$0 = f(0)$$

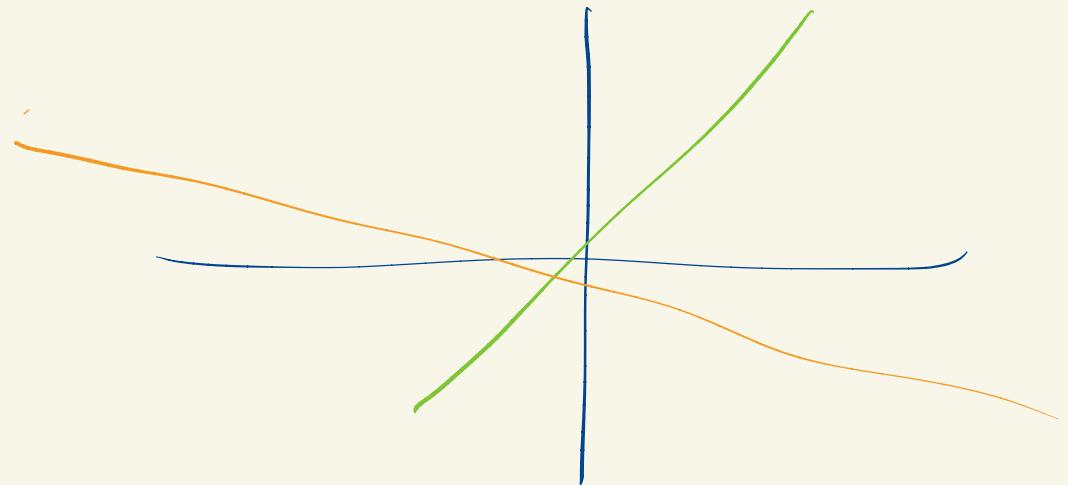
If  $f$  is linear then  $f(0) = 0$ .

$$f(x) = mx + b$$

$$\underbrace{f(x) = mx}$$

$$f(0) = b$$

This is linear.



A function  $f(x)$  of the form

$$f(x) = c^T x + b \quad \text{where } b \in \mathbb{R}$$

is called an affine function.

Linear functions:

$$f(\alpha x + \beta y) = \alpha f(x) + \beta f(y)$$

Affine functions

$$f(\alpha x + \beta y) = \alpha f(x) + \beta f(y)$$

but only if  $\alpha + \beta = 1$ .

(limited superposition)

↳ see text.

$$\boxed{x = (x_1, x_2)}$$

$$f(x_1, x_2) = x_1 \cdot e^{x_2}$$

$$f(3, 0) = 3 \cdot e^0 = 3$$

$f(3.1, -0.2)$  is close to 3

Can we approximate this value using only the skills of 4<sup>th</sup> grade?

$$\frac{\partial f}{\partial x_1} = e^{x_2}$$

$$\frac{\partial f}{\partial x_2} = x_1 e^{x_2}$$

at  $(3, 0)$

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix}$$

$$\nabla f = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

We can approximate  $f(x_1, x_2)$  for  $(x_1, x_2)$  near

as follows:

$$\hat{f}(x_1, x_2) = f(3, 0) + \frac{\partial f}{\partial x_1} \text{ at } (3, 0)(x_1 - 3) + \frac{\partial f}{\partial x_2} \text{ at } (3, 0)(x_2 - 0)$$

$\frac{\partial f}{\partial x_1} \text{ at } (3, 0)$        $\downarrow$        $(3, 0)$   
 $\downarrow$        $\downarrow$        $\uparrow$   
 $\hat{f}(x_1, x_2) = f(3, 0) + \frac{\partial f}{\partial x_1} \text{ at } (3, 0)(x_1 - 3) + \frac{\partial f}{\partial x_2} \text{ at } (3, 0)(x_2 - 0)$   
 $\uparrow$        $\uparrow$   
 $f(3, 0)$        $\frac{\partial f}{\partial x_2} \text{ at } (3, 0)$   
 $(3, 0)$        $\downarrow$

$$\hat{f}(3, 2, -0.1) = 3 + 1(0.2) + 3(-0.1)$$

$$= 3 + 0.2 - 0.3$$

$$= 3 - 0.1 = 2.9$$

$$f(3.2, -0.1) = 2.895$$

$$f(1, -10) = 4.5 \times 10^{-5} \quad (\text{close to } 0)$$

$$\hat{f}(1, -10) = 31$$

$$w = (3, 0)$$

$$\hat{f}(x) = f(3, 0) + \underbrace{(\nabla f)^T}_{\text{evaluated}} (x - (3, 0))$$

$$x = (x_1, x_2)$$

evaluated

$$\text{at } (3, 0)$$

linear  
approximation

$$\hat{f}(x) = f(w) + (\nabla f)^T (x - w)$$

evaluated at  $w$

an approximation of  $f$  near  $w_0$ .  $\nabla f(w)$

$$\hat{f}(x) = \left[ f(w) - \nabla f^T w \right] + (\nabla f)^T x$$

$b$

This is an affine function of  $x$