

# **Regression and Simulation Methods**

## **Week 6: Generalised Linear Models**

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# What has been covered over the last week?

- The Poisson distribution
- Logistic regression
- Exponential family
- GLMs



# Exponential Family

# Definition

$$f(y; \theta) = \exp(a(y)b(\theta) + c(\theta) + d(y))$$

$a(y)$  - Canonical link function

$b(\theta)$  - natural parameter



# Expectation and Variance

**Proposition 6.2.** For a random variable  $Y$  in exponential family form (6.3),

$$(i) \quad \mathbb{E}(a(Y)) = \frac{-c'(\theta)}{b'(\theta)}$$

$$(ii) \quad \text{var}(a(Y)) = \frac{b''(\theta)c'(\theta) - c''(\theta)b'(\theta)}{[b'(\theta)]^3}$$

# Example

## The Gaussian distribution

The (univariate) Gaussian density can be written as follows (where the underlying measure is Lebesgue measure):

$$p(x \mid \mu, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp \left\{ -\frac{1}{2\sigma^2} (x - \mu)^2 \right\} \quad (8.19)$$

$$= \frac{1}{\sqrt{2\pi}} \exp \left\{ \frac{\mu}{\sigma^2} x - \frac{1}{2\sigma^2} x^2 - \frac{1}{2\sigma^2} \mu^2 - \log \sigma \right\}. \quad (8.20)$$

This is in the exponential family form, with:

$$\text{b(theta)} \quad \eta = \begin{bmatrix} \mu/\sigma^2 \\ -1/2\sigma^2 \end{bmatrix} \quad (8.21)$$

$$\text{a(y)} \quad T(x) = \begin{bmatrix} x \\ x^2 \end{bmatrix} \quad (8.22)$$

$$\text{c(theta)} \quad A(\eta) = \frac{\mu^2}{2\sigma^2} + \log \sigma = -\frac{\eta_1^2}{4\eta_2} - \frac{1}{2} \log(-2\eta_2) \quad (8.23)$$

$$\text{NOT d(y)} \quad h(x) = \frac{1}{\sqrt{2\pi}}. \quad (8.24)$$



# GLMs

# Definition

## Definition 2: Generalised Linear Model

A generalised linear model (GLM) consists of the following three components:

1. A (common) distribution from the exponential family for the independent response variables  $Y_1, \dots, Y_n$ .
2. A linear predictor

$$\eta_i = \beta_0 + \beta_1 x_{i1} + \dots + \beta_p x_{ip}.$$

3. A differentiable, strictly monotone link function such that,

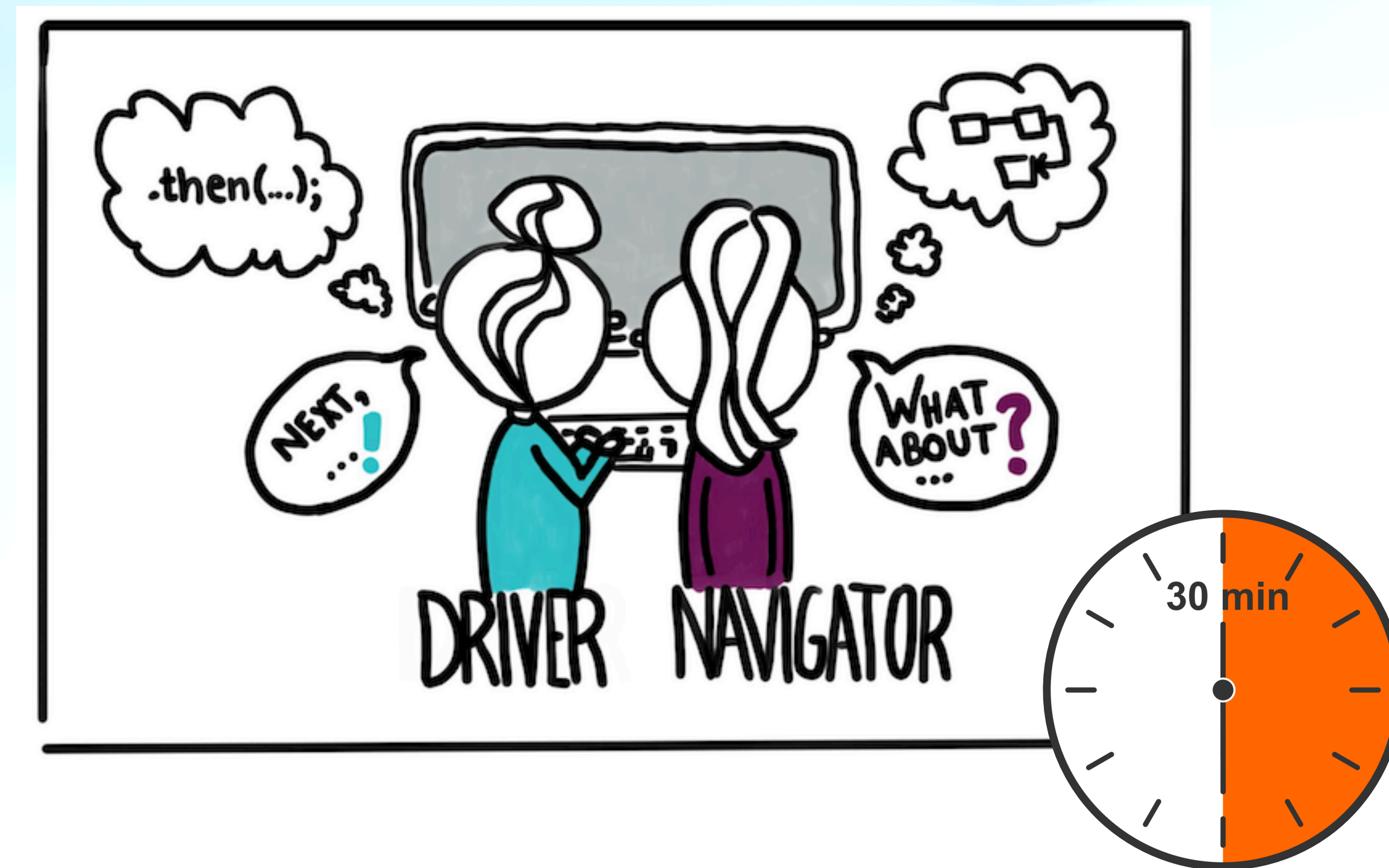
$$g(\mu_i) = \eta_i,$$

where  $\mu_i = \mathbb{E}[Y_i]$



# Rest of the tutorial...

- In pairs work on the sixth notebook. Paired programming will continue!





# **Rest of the week...**

**Once you have finished the R-Script go ahead and answer 6-1 through 6-4 from the SMSTC resources.**



# Additional Questions

1. The gamma pdf of a random variable  $Y$  is given by:

$$f(y) = (s^a \Gamma(a))^{-1} y^{a-1} e^{-y/s}$$

where  $y \geq 0$ ,  $s$  is the *scale* parameter,  $a$  the *shape* parameter, and  $\Gamma$  the gamma function. (If  $a$  is integer then  $\Gamma(a) = (a - 1)!$ )

- (a) Reparameterise this pdf by setting  $a = 1/\phi$  and  $s = \mu\phi$ , and hence show that it is a member of the exponential family.
- (b) Deduce that the canonical link for the gamma is  $\theta = \frac{1}{\mu} = \eta = \mathbf{X}\beta$ .
- (c) Using the result above, compute  $b''(\theta)$ .
- (d) Compute  $V[Y]$  (hint: you can use Proposition 6.2).
- (e) How does this last result help if you want to assess whether data are best modelled as a normal, overdispersed Poisson or gamma distribution?