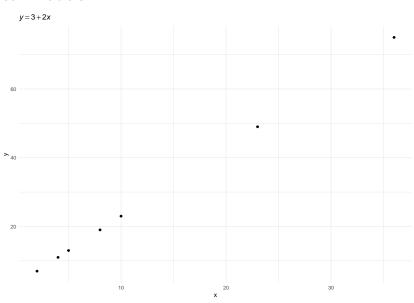
An introduction to GAM(M)s

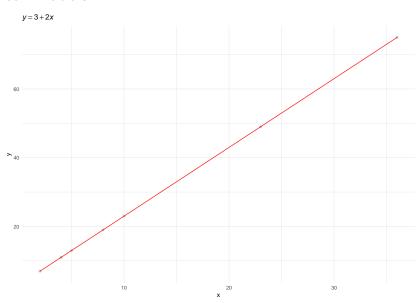
Stefano Coretta

12/07/2018

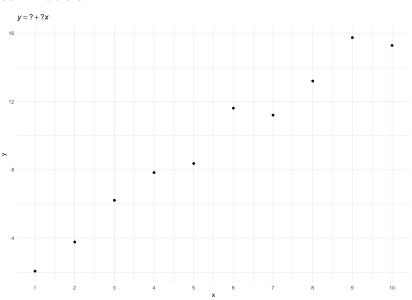
Time travel...

$$y = 3 + 2x$$
where $x = (2, 4, 5, 8, 10, 23, 36)$

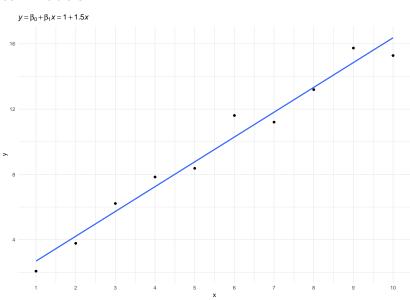


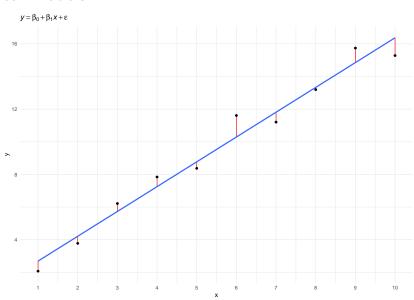


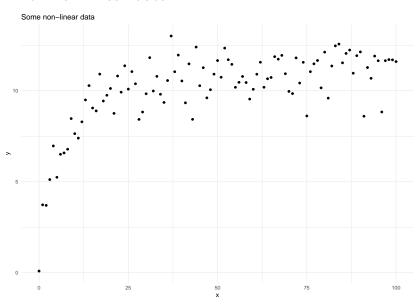
- \blacktriangleright In science, we have x and y...
- ▶ for example, vowel duration and VOT

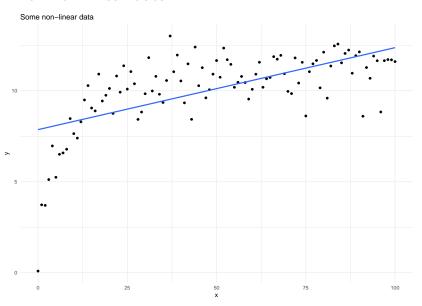


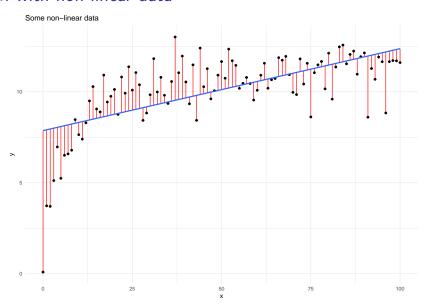
- ▶ The formula: $y = \beta_0 + \beta_1 x$
 - \triangleright β_0 is the **intercept**
 - $ightharpoonup \beta_1$ is the **slope**
- ▶ We know x and y
 - we need to estimate β_0 , $\beta_1 = \hat{\beta}_0$, $\hat{\beta}_1$
- We can add more predictors
 - $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + ... + \beta_n x_n$
- ▶ $lm(y \sim x, data)$ ('y as a function of x')





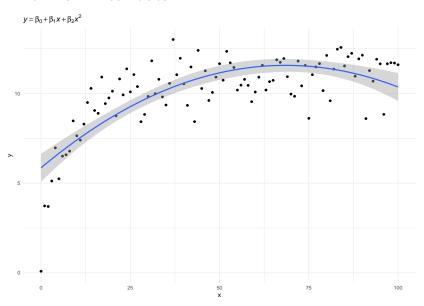


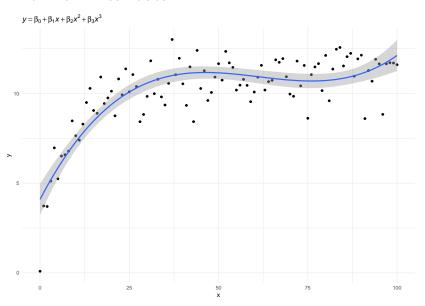


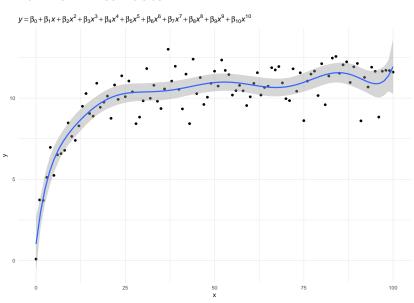


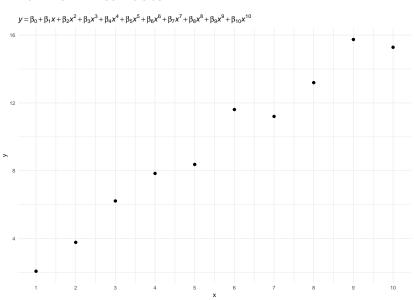
How to account for non-linearity in a linear model?

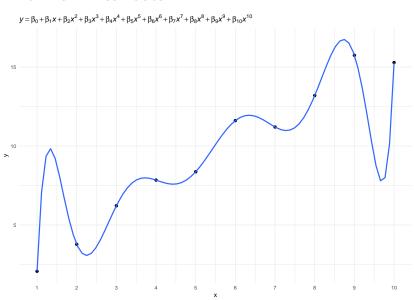
- Use higher-degree polynomials
 - quadratic: $y = \beta_0 + \beta_1 x + \beta_2 x^2$
 - cubic: $y = \beta_0 + \beta_1 x + \beta_2 x^2 + \beta_3 x^3$
 - *n*th: $y = \beta_0 + \beta_1 x + \beta_2 x^2 + \beta_3 x^3 + ... + \beta_n x^n$











Generalised additive models

- ► Genrealised Additive Models
- $ightharpoonup y = f(x) + \epsilon$
 - f(x) = 'some function of x' (or *smooth function*)

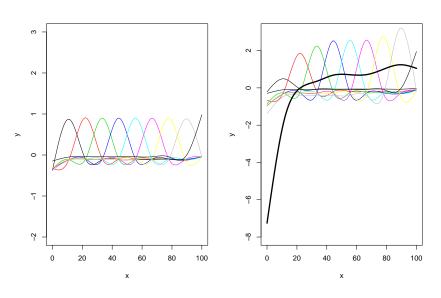
Smooth terms

- ► LMs have parametric terms
 - $\triangleright \beta_n x_n \text{ (x in R)}$
 - linear effects
- GAMs add (non-parametric) smooth terms (or simply smooths, also smoothers)
 - ightharpoonup f(x), s(x) in R
 - non-linear effects
- ▶ gam(y ~ s(x), data), 'y as some function of x'

Smoothing splines, basis, basis functions

- smooths in GAMs are smoothing splines
 - splines are defined piecewise with a set of polynomials
- the set of polynomials is called a basis
 - the basis is composed of basis functions (the polynomials)
- a spline is the sum of the products of each basis function and its coefficient

Basis functions



Smoothing parameter

- 'wiggliness' is related to number of basis functions
 - more basis functions, more wiggliness (less smoothing)
- the smoothing parameter penalises wiggliness
 - high values = less wiggliness (more smoothing)
 - estimated from the data

Smoothing splines

- there are several kinds of splines
 - each with their own basis functions
- most common
 - thin plate regression splines
 - cubic regression splines
- for more info, run ?smooth.terms

A simple GAM

```
simple <- gam(y ~ s(x, bs = "cr", k = 10), data = sim_nl_a)
summary(simple)
##</pre>
```

```
## Family: gaussian
```

```
## Link function: identity
```

```
## Formula:
```

##

##

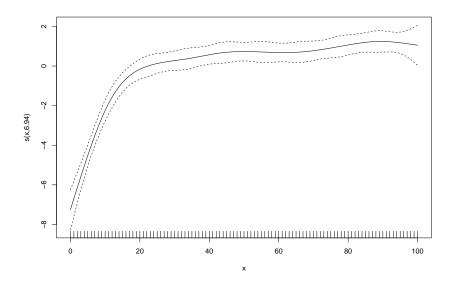
```
## y ~ s(x, bs = "cr", k = 10)
```

```
## Parametric coefficients:
```

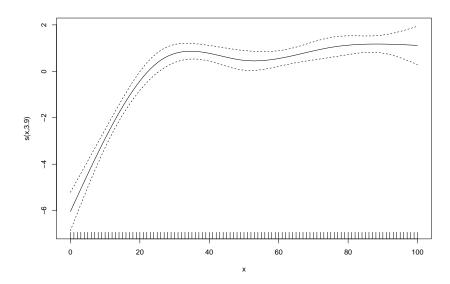
Estimate Std. Error t value Pr(>|t|)

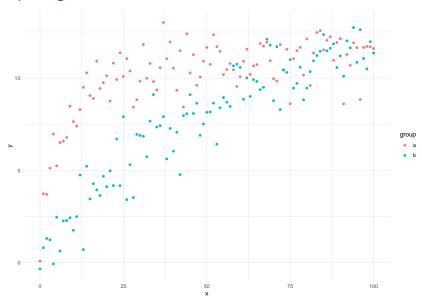
26 / 36

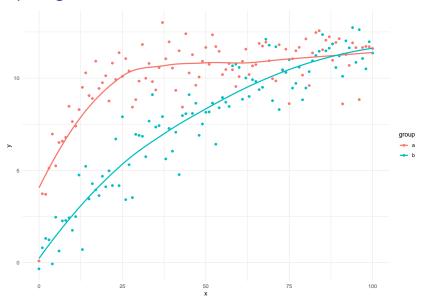
A simple GAM



A simple GAM







ordered by-variables

```
compare <- gam(</pre>
    group +
    s(x, bs = "cr", k = 5) +
    s(x, bs = "cr", k = 5, by = group),
  data = sim_nl
```

- to use ordered by-variables
 - change factor to ordered factor
 - change factor contrast to treatment contrast
 - (contr.treatment)
 - the default in ordered factors is contr.poly, this won't work
 - include factor as parametric term
 - include a reference smooth and a difference smooth with the by-variable

```
sim_nl <- sim_nl %>%
  mutate(group = ordered(group, levels = c("a", "b")))
contrasts(sim_nl$group) <- "contr.treatment"</pre>
```

```
compare <- gam(</pre>
    group +
    s(x, bs = "cr", k = 5) +
    s(x, bs = "cr", k = 5, by = group),
  data = sim_nl
```

```
##
## Family: gaussian
## Link function: identity
##
## Formula:
## y ~ group + s(x, bs = "cr", k = 5) + s(x, bs = "cr", k = 5)
##
## Parametric coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 10.1165 0.1096 92.34 <2e-16 ***
## groupb -2.4947 0.1549 -16.10 <2e-16 ***
```

35 / 36

Compare levels

