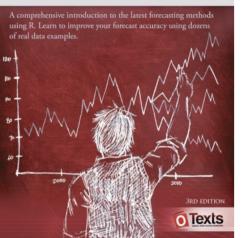
Rob J Hyndman George Athanasopoulos

FORECASTING PRINCIPLES AND PRACTICE



7. Time series regression models

7.7 Nonlinear regression

OTexts.org/fpp3/

Nonlinear regression

A log-log functional form

$$\log y = \beta_0 + \beta_1 \log x + \varepsilon$$

where β_1 is interpreted as an elasticity (the average percentage change in y resulting from a 1% increase in x).

2

Nonlinear regression

A log-log functional form

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where β_1 is interpreted as an elasticity (the average percentage change in y resulting from a 1% increase in x).

- alternative specifications: log-linear, linear-log.
- use log(x + 1) if required.

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

where f is a non-linear function.

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

where *f* is a non-linear function.

For piecewise linear let $x_1 = x$ and

$$x_2 = (x - c)_+ = \begin{cases} 0 & \text{if } x < c \\ x - c & \text{if } x \ge c. \end{cases}$$

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■ In general, linear regression splines

$$x_1 = x$$
 $x_2 = (x - c_1)_+$... $x_k = (x - c_{k-1})_+$

where c_1, \ldots, c_{k-1} are knots.

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■ In general, linear regression splines

$$x_1 = x$$
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where c_1, \ldots, c_{k-1} are knots.

- Need to select knots: can be difficult and arbitrary.
- Automatic knot selection algorithms very slow.
- Using piecewise cubics achieves a smoother result.

Warning: better fit but forecasting outside the range of the historical data is even more unreliable.

Nonlinear trends

Piecewise linear trend with bend at au

$$x_{1,t} = t$$

$$x_{2,t} = \begin{cases} 0 & t < \tau \\ (t - \tau) & t \ge \tau \end{cases}$$

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Quadratic or higher order trend

$$x_{1,t} = t, \quad x_{2,t} = t^2, \quad \dots$$

Nonlinear trends

Piecewise linear trend with bend at au

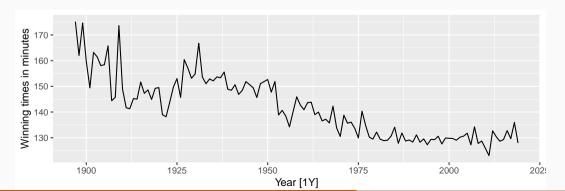
$$x_{1,t} = t$$

$$x_{2,t} = \begin{cases} 0 & t < \tau \\ (t - \tau) & t \ge \tau \end{cases}$$

Quadratic or higher order trend

$$x_{1,t} = t$$
, $x_{2,t} = t^2$, ...
NOT RECOMMENDED!

```
marathon <- boston_marathon |>
  filter(Event == "Men's open division") |>
  select(-Event) |>
  mutate(Minutes = as.numeric(Time) / 60)
marathon |> autoplot(Minutes) + labs(y = "Winning times in minutes")
```



```
fit_trends <- marathon |>
  model(
    # Linear trend
  linear = TSLM(Minutes ~ trend()),
    # Exponential trend
  exponential = TSLM(log(Minutes) ~ trend()),
    # Piecewise linear trend
  piecewise = TSLM(Minutes ~ trend(knots = c(1940, 1980)))
)
```

fit_trends

```
## # A mable: 1 x 3
## linear exponential piecewise
## <model> <model> <model>
## 1 <TSLM> <TSLM> <TSLM>
```

```
fit_trends |>
  forecast(h = 10) |>
  autoplot(marathon)
```

