Probabilistic forecasts for anomaly detection

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- 1 Australian PBS data
- 2 Probabilistic forecasts
- 3 Extreme log scores
- 4 Using the fable and weird packages
- 5 Online anomaly detection

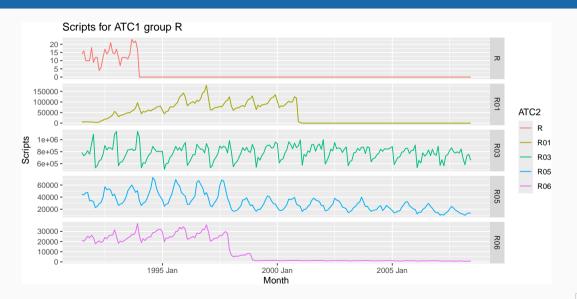
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Australian PBS data

```
pbs
```

```
# A tsibble: 17,016 x 4 [1M]
# Kev:
             ATC1, ATC2 [84]
  ATC1
        ATC2
                  Month Scripts
   <chr> <chr>
                  <mth>
                          <dbl>
 1 A
         A01
              1991 Jul
                          22615
 2 A
         A01
               1991 Aug
                          20443
 3 A
         A01
               1991 Sep
                          21389
4 A
         A01
               1991 Oct
                          23746
 5 A
         A01
               1991 Nov
                          23477
 6 A
         A01
               1991 Dec
                          26316
 7 A
         A01
               1992 Jan
                          22041
8 A
         A01
               1992 Feb
                          16393
9 A
         A01
               1992 Mar
                          17207
10 A
         A01
               1992 Apr
                          18847
# i 17,006 more rows
```

Australian PBS data



Main idea

- Look at one-step forecast distributions and compute the anomaly score = log probability density of the next observation.
- High anomaly scores indicate potential anomalies.
- Fit a Generalized Pareto Distribution to the largest anomaly scores.
- Estimate the probability of each observation being an anomaly.

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Probabilistic forecasts

Describe probabilistic forecasts

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Extreme log scores

Explain log scores and EVT approach to finding anomalies ala lookout

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Using the fable and weird packages

Demonstrate using weird package with (a) univariate models for tourism data; and (b) univariate models for age-specific time series from French mortality.

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Online anomaly detection

Need to compute one-step forecast distributions using prior data. Approximately the same as N(y-hat, sigma^2) using in-sample fit. More accurate using stretch tsibble with one-step forecasts.