

# Newton Raphson for Optimization

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# Poisson Model Example

Example A in Section 8.4 of Rice

The National Institute of Science and Technology did a study where they wanted to develop measurement standards for asbestos concentration. Asbestos dissolved in water was spread on a filter, and 3-mm diameter punches were taken from the filter and mounted on a transmission electron microscope. An operator counted the number of fibers in each of 23 grid squares. For the sake of illustration I am using just the first 5 observations here.

Let  $X_i$  be the number of fibers of asbestos found in square number  $i$ .

Model:  $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Poisson}(\lambda)$

# Log-likelihood and its derivatives

$$f(x_i | \lambda) = e^{-\lambda} \frac{\lambda^{x_i}}{x_i!}$$

$$\ell(\lambda | x_1, \dots, x_n) = \dots = -n\lambda + \log(\lambda) \sum_{i=1}^n x_i - \sum_{i=1}^n \log(x_i)$$

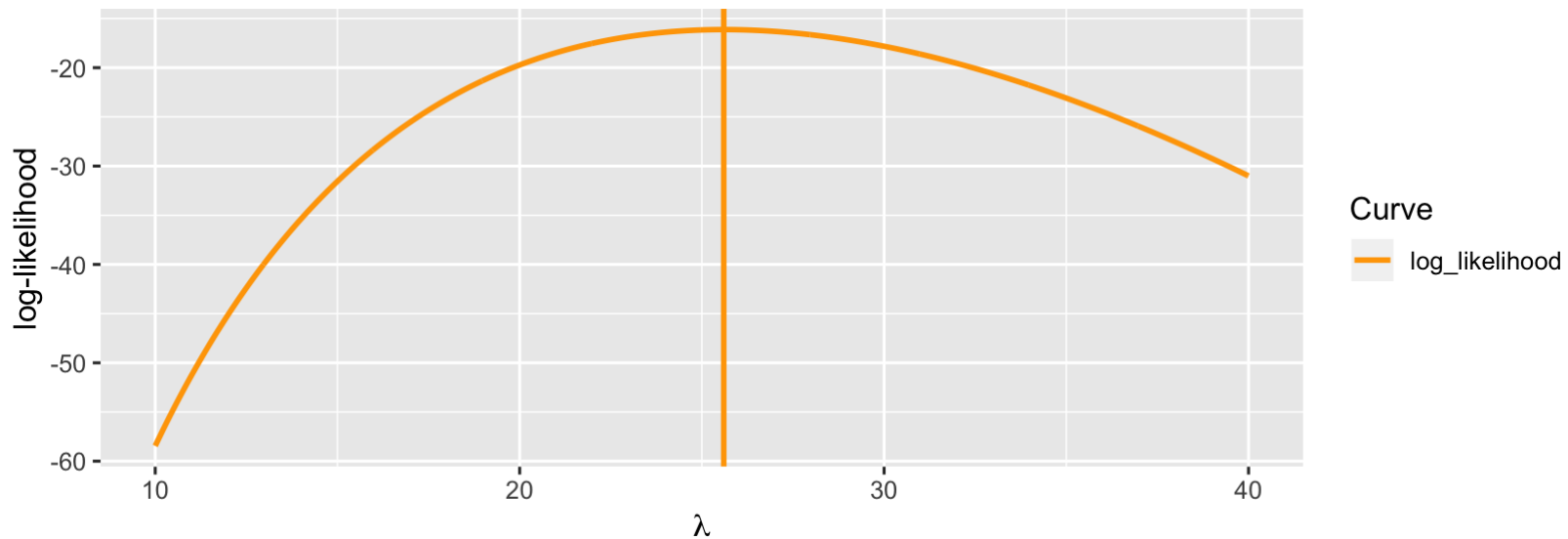
$$\frac{d}{d\lambda} \ell(\lambda | x_1, \dots, x_n) = \frac{1}{\lambda} \sum_{i=1}^n x_i - n$$

$$\frac{d^2}{d\lambda^2} \ell(\lambda | x_1, \dots, x_n) = \frac{-1}{\lambda^2} \sum_{i=1}^n x_i$$

We saw previously that  $\hat{\lambda}_{MLE} = \frac{1}{n} \sum_{i=1}^n x_i$

In our example,  $\hat{\lambda}_{MLE} = 25.6$

# Log-likelihood function, MLE



The vertical orange line is at the MLE.

...But what if we couldn't solve for the MLE directly?

# Taylor Series Approximation to $L$

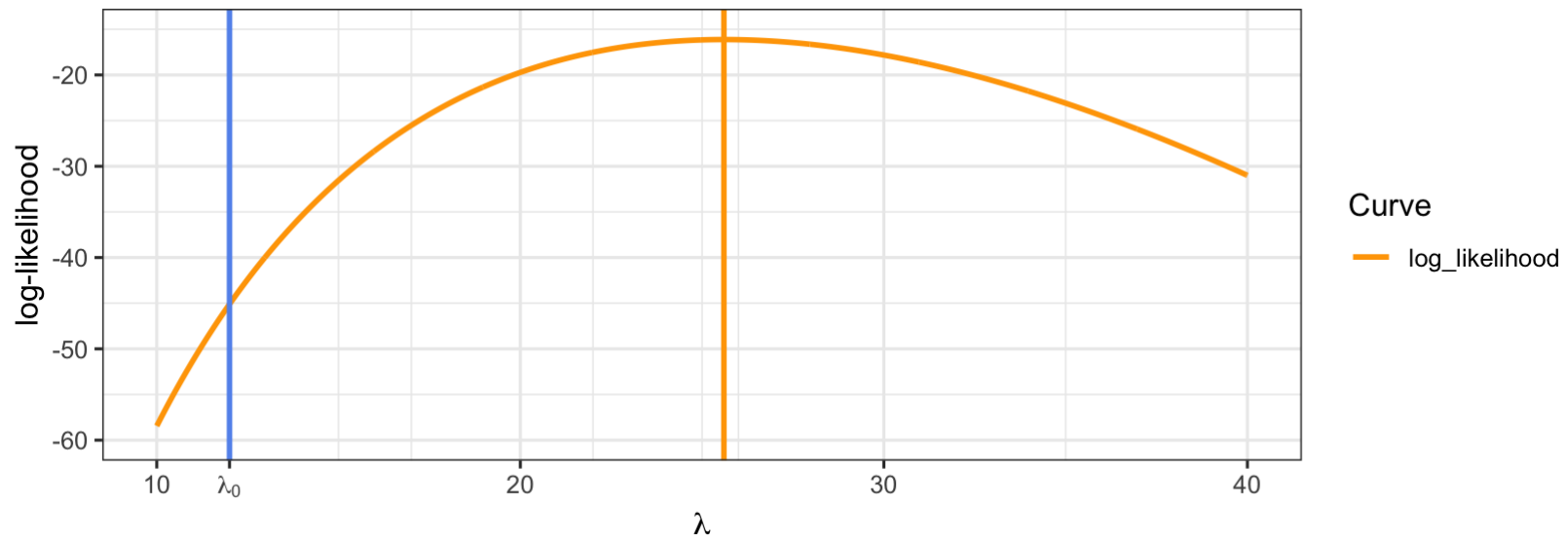
Pick a value  $\lambda_0$ . The second-order Taylor Series approximation to  $L(\lambda|x_1, \dots, x_n)$  around  $\lambda_0$  is

$$P_2(\lambda) = L(\lambda_0|x_1, \dots, x_n) + \frac{d}{d\lambda}L(\lambda_0|x_1, \dots, x_n)(\lambda - \lambda_0) + \frac{1}{2} \frac{d^2}{d\lambda^2}L(\lambda_0|x_1, \dots, x_n)(\lambda - \lambda_0)^2$$

The maximum of  $P_2(\lambda)$  is at  $\lambda_1 = \lambda_0 - \frac{\frac{d}{d\lambda}L(\lambda_0|x_1, \dots, x_n)}{\frac{d^2}{d\lambda^2}L(\lambda_0|x_1, \dots, x_n)}$

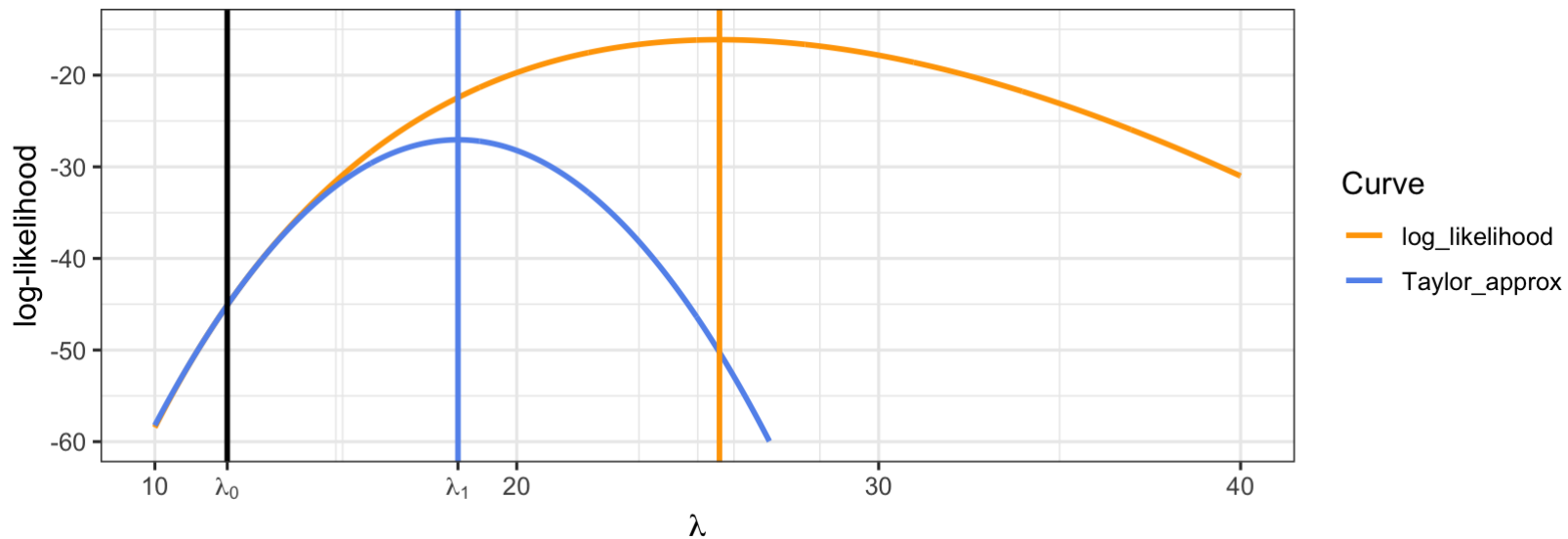
Now repeat, but centering the Taylor Series approximation at  $\lambda_1$ .

# Pick $\lambda_0$



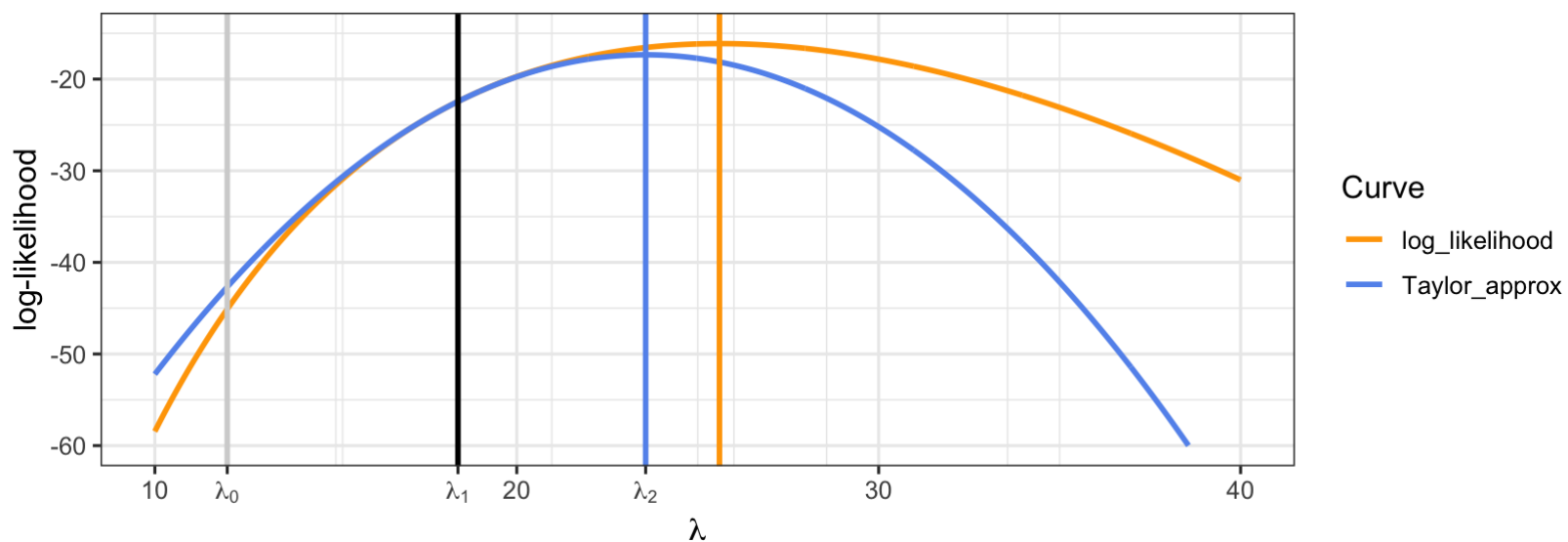
$$\lambda_0 = 12$$

# Approximate $L$ around $\lambda_0$ , get $\lambda_1$



$$\lambda_0 = 12, \lambda_1 = \lambda_0 - \frac{\frac{d}{d\lambda}L(\lambda_0|x_1, \dots, x_n)}{\frac{d^2}{d\lambda^2}L(\lambda_0|x_1, \dots, x_n)} = 18.375$$

# Approximate $L$ around $\lambda_1$ , get $\lambda_2$

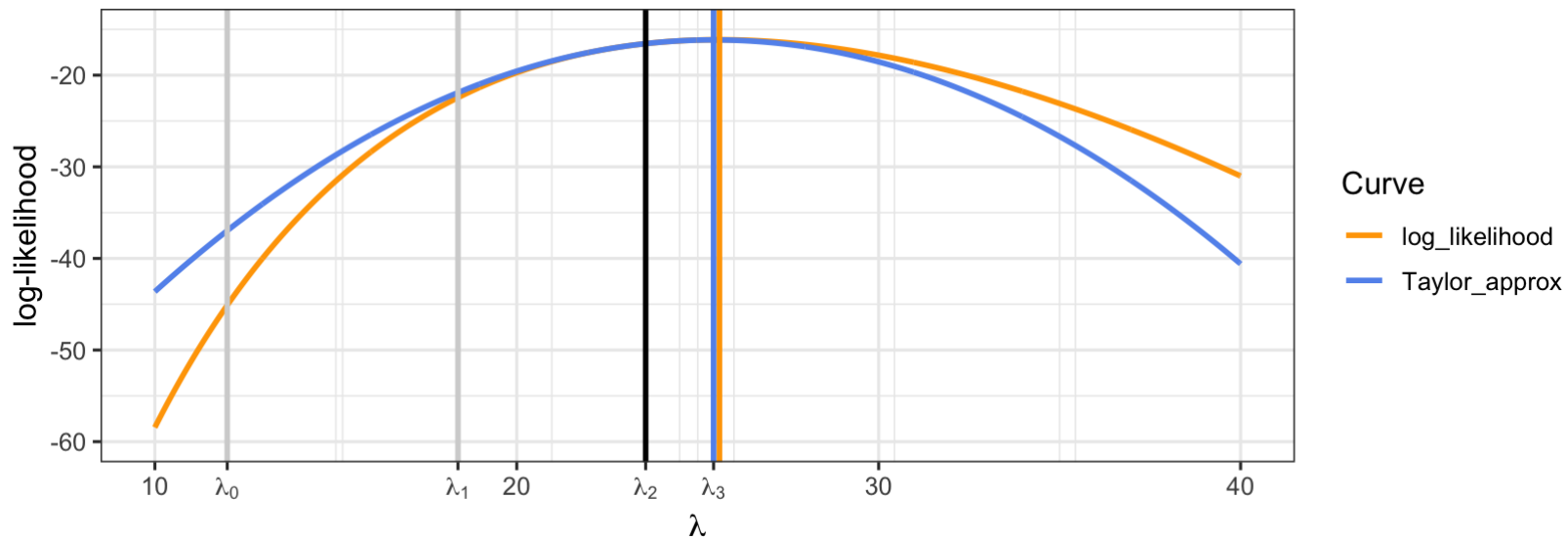


$$\lambda_0 = 12, \lambda_1 = 18.375,$$

$$\lambda_2 = \lambda_1 - \frac{\frac{d}{d\lambda} L(\lambda_1 | x_1, \dots, x_n)}{\frac{d^2}{d\lambda^2} L(\lambda_1 | x_1, \dots, x_n)} = 23.561$$



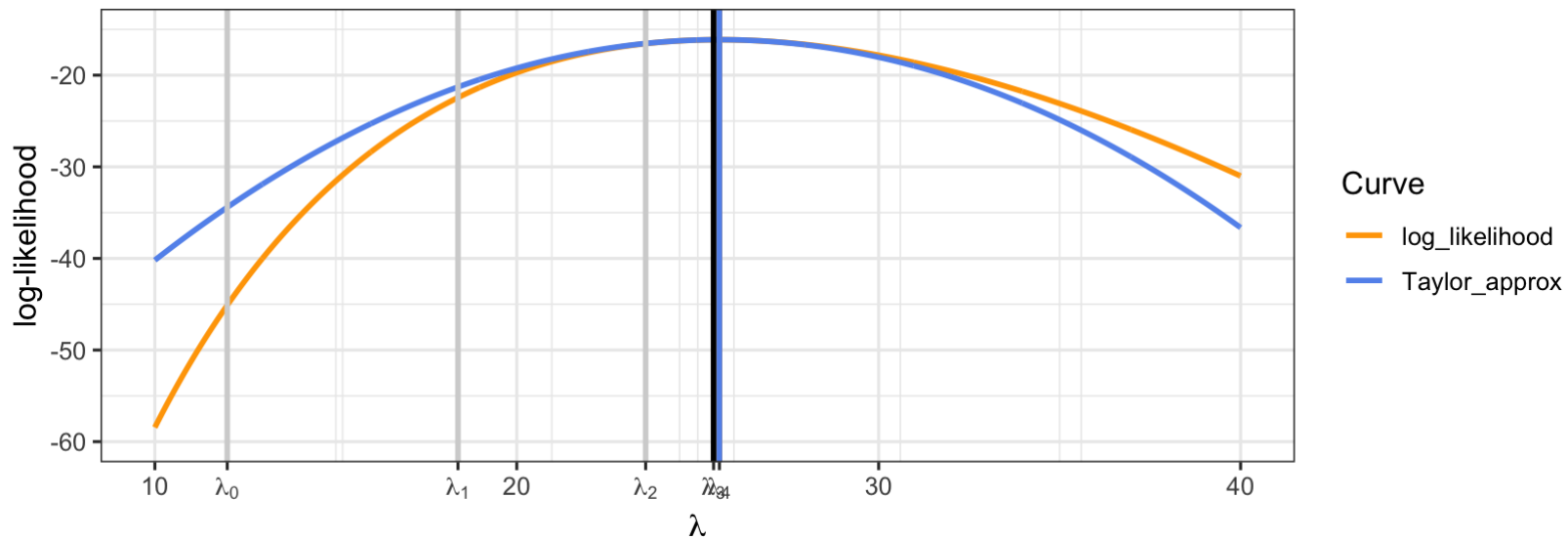
# Approximate $L$ around $\lambda_2$ , get $\lambda_3$



$$\lambda_0 = 12, \lambda_1 = 18.375, \lambda_2 = 23.561$$

$$\lambda_3 = \lambda_2 - \frac{\frac{d}{d\lambda} L(\lambda_2 | x_1, \dots, x_n)}{\frac{d^2}{d\lambda^2} L(\lambda_2 | x_1, \dots, x_n)} = 25.438$$

# Approximate $L$ around $\lambda_3$ , get $\lambda_4$



$$\lambda_0 = 12, \lambda_1 = 18.375, \lambda_2 = 23.561, \lambda_3 = 25.438$$

$$\lambda_4 = \lambda_3 - \frac{\frac{d}{d\lambda} L(\lambda_3 | x_1, \dots, x_n)}{\frac{d^2}{d\lambda^2} L(\lambda_3 | x_1, \dots, x_n)} = 25.599$$