

# On the Gaussian Mixture Representation of the Laplace Distribution with Location and Scale Parameters and Applications.

Seth Agee

Tennessee Technological University

## Objectives

Objectives for today:

- An overview of the Laplace distribution.
- Showing that the Laplace distribution with location and scale parameters has a Gaussian mixture representation.
- An introduction to an asymmetric Laplace distribution.
- Fitting speed data with the Laplace and Asymmetric Laplace distributions.

## The Laplace Distribution

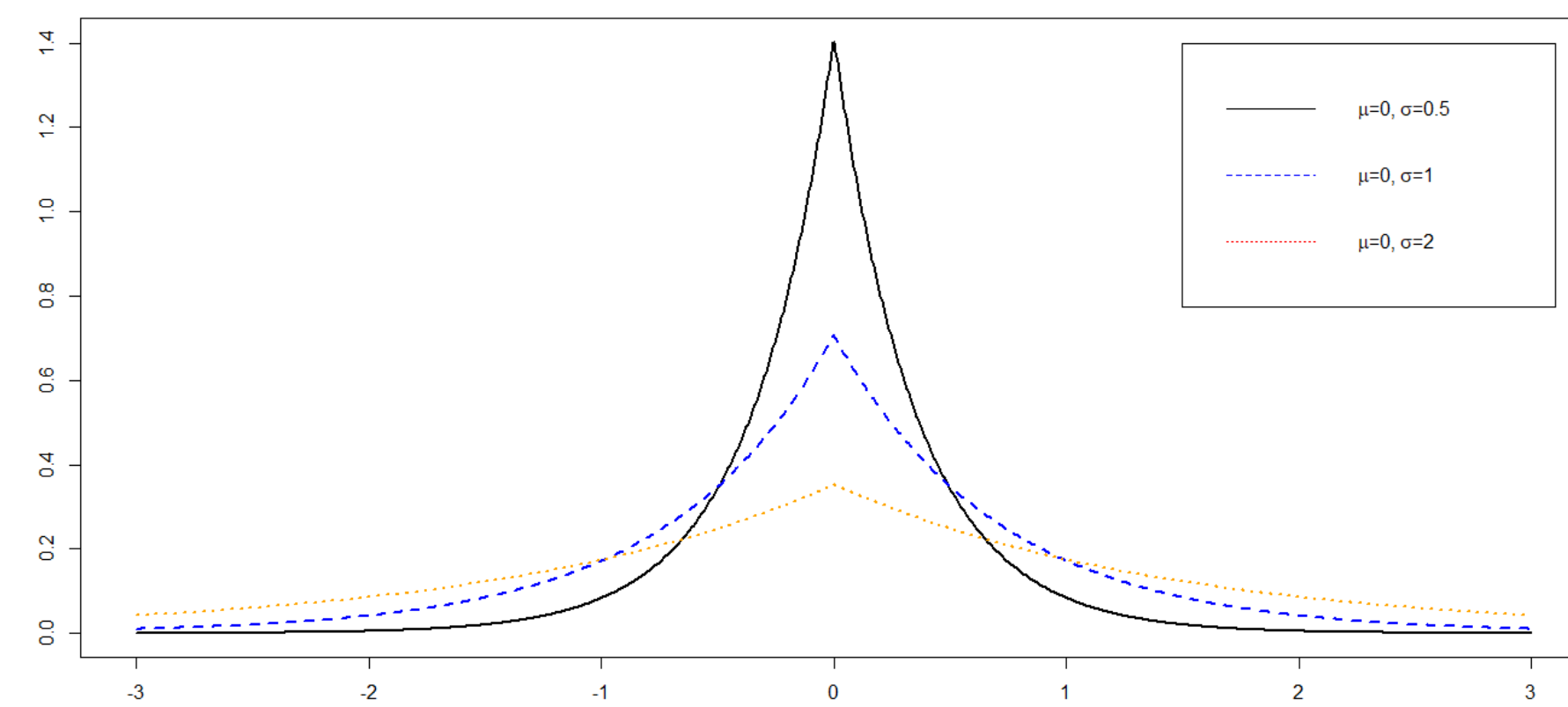


Figure 1: Three Laplace Density Curves with Varying Scale Parameters.

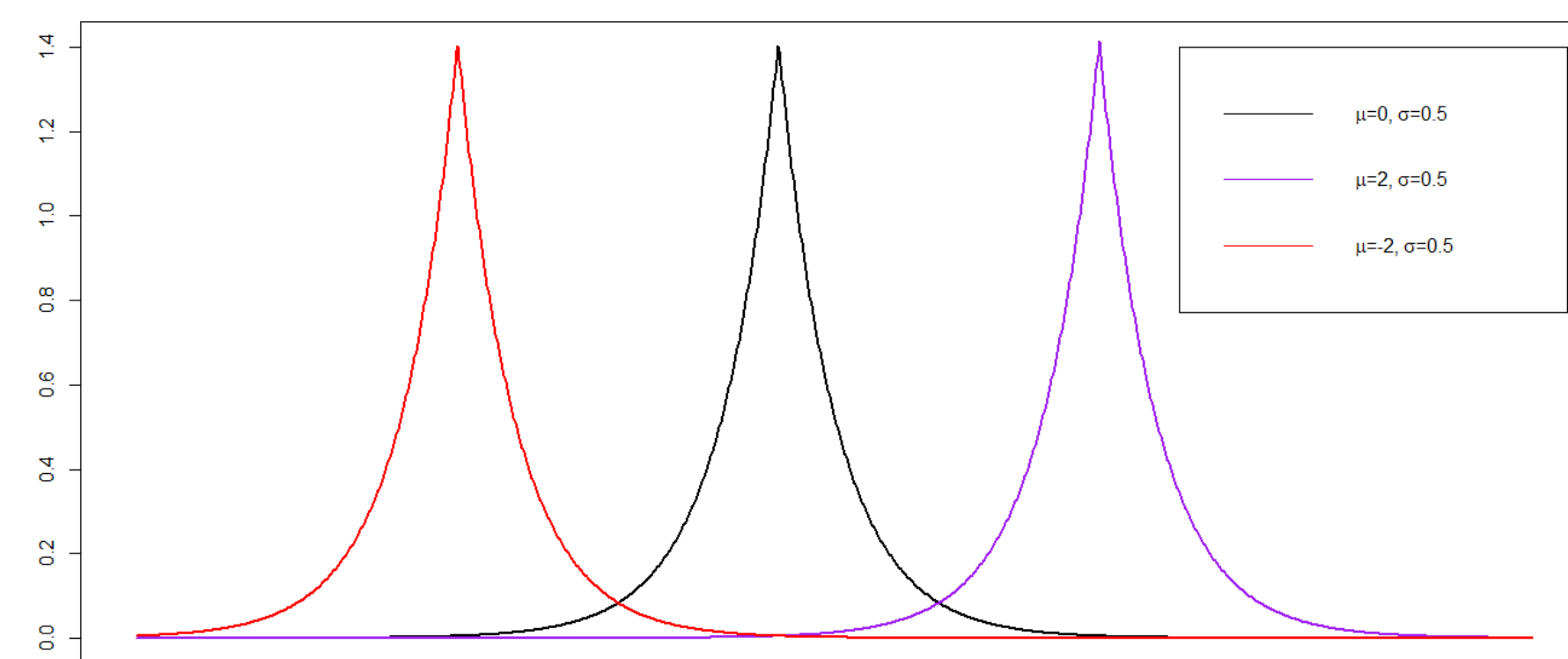


Figure 2: Three Laplace Density Curves with Varying Location Parameters.

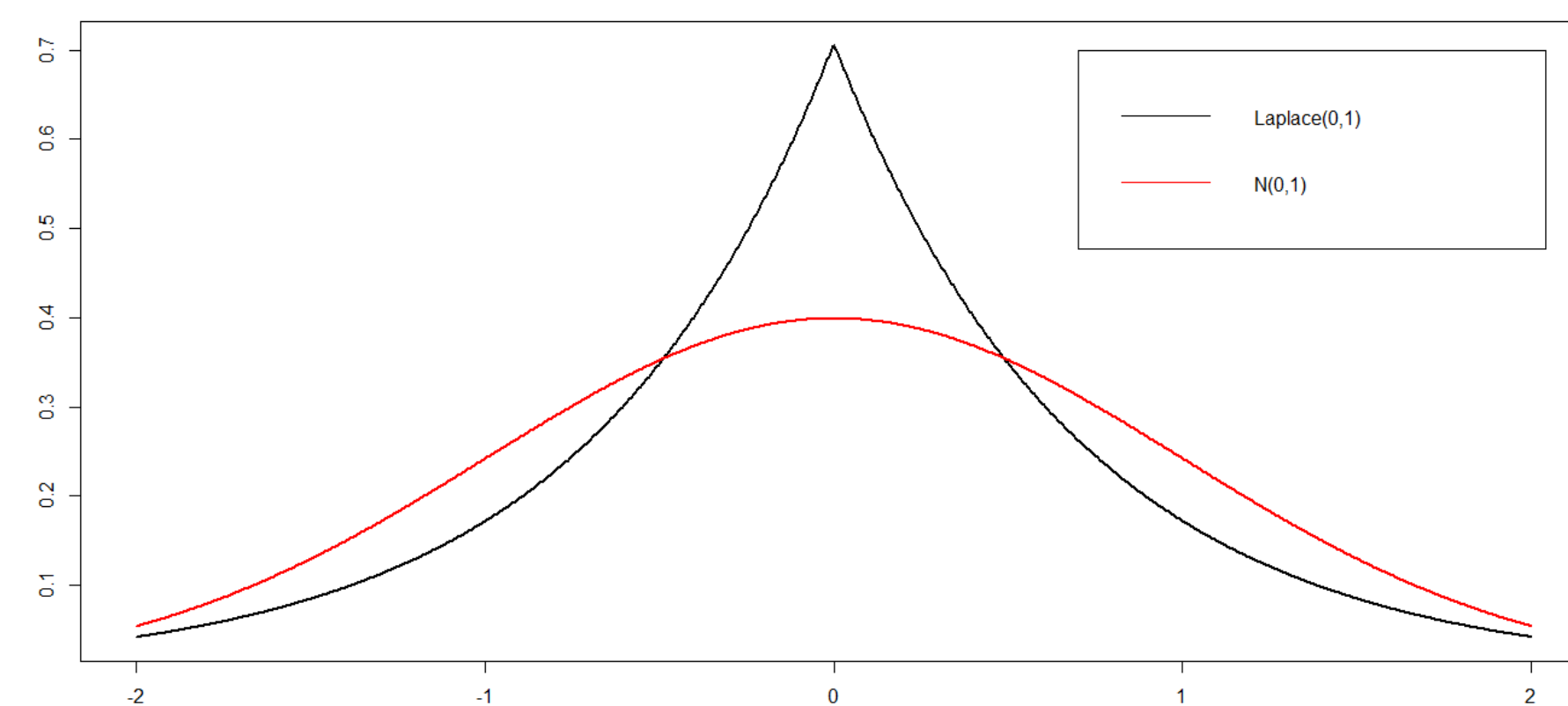


Figure 3: Laplace Distribution and Standard Normal Distribution.

- Very pronounced peak at the center.
- Sometimes called the double Exponential distribution.
- Can be thought of as putting a positive and negative Exponential distribution back-to-back.
- Has cumulative distribution function (CDF) (Kotz, Samuel and Kozubowski, Tomasz and Podgorski, Krzysztof, 2001)

$$F(x) = \begin{cases} \frac{1}{2}e^{-(x-\mu)/\sigma} & x \leq \mu \\ 1 - \frac{1}{2}e^{-(x-\mu)/\sigma} & x \geq \mu \end{cases}$$

## Notation

From now on we use the following notation:

- Let  $\mathcal{E}$  be an Exponential random variable with mean 1.
- Let  $Z$  be a standard Normal random variable,  $N(0,1)$ .
- Let  $\mathcal{L}$  be a Laplace random variable.
- Let  $S$  be a random sign taking values -1 and +1 with equal probabilities 1/2.

## The Representation $\mathcal{L} \sim \mu + \sigma S\mathcal{E}$

First, note that  $P(\mathcal{E} \leq x) = 1 - e^{-x}$  when  $x \geq 0$  or 0 when  $x < 0$ . Now consider

$$\begin{aligned} P(\mu + \sigma S\mathcal{E} \leq x) &= P(S\mathcal{E} \leq \frac{x-\mu}{\sigma}) \\ &= P(S = -1)P(\mathcal{E} \geq \frac{-x+\mu}{\sigma}) + P(S = 1)P(\mathcal{E} \leq \frac{x-\mu}{\sigma}). \end{aligned}$$

- If  $x \leq \mu$ , then  $P(\mu + \sigma S\mathcal{E} \leq x) = \frac{1}{2}e^{(x-\mu)/\sigma}$ .
- If  $x > \mu$ , then  $P(\mu + \sigma S\mathcal{E} \leq x) = 1 - \frac{1}{2}e^{-(x-\mu)/\sigma}$ .
- This gives us exactly the CDF of the Laplace distribution, and so we get  $\mathcal{L} \sim \mu + \sigma S\mathcal{E}$ .

## The Laplace Distribution has a Gaussian Mixture Representation

First, note that moment generating functions (MGFs) of probability distributions are unique. Also note that the MGF of  $Z$  is denoted  $M_Z$  and the MGF of  $\mathcal{E}$  is denoted by  $M_{\mathcal{E}}$ , and that  $M_Z = e^{t^2/2}$  and  $M_{\mathcal{E}} = \frac{1}{1-t}$  for  $t < 1$ . Let  $M_{\mu+\sigma S\mathcal{E}}$  and  $M_{\mu+\sigma\sqrt{2}\mathcal{E}Z}$  be the MGFs of  $\mu + \sigma S\mathcal{E}$  and  $\mu + \sigma\sqrt{2}\mathcal{E}Z$  respectively. Showing these are equal will show that  $\mathcal{L} \sim \mu + \sigma\sqrt{2}\mathcal{E}Z$ .

For  $\mu + \sigma\sqrt{2}\mathcal{E}Z$  we get

$$\begin{aligned} M_{\mu+\sigma\sqrt{2}\mathcal{E}Z} &= E[e^{t(\mu+\sigma\sqrt{2}\mathcal{E}Z)}] = e^{\mu t} E[e^{t\sigma\sqrt{2}\mathcal{E}Z}] = e^{\mu t} E[E[e^{t\sigma\sqrt{2}\mathcal{E}Z} | \mathcal{E}]] \\ &= e^{\mu t} E[M_Z(t\sigma\sqrt{2}\mathcal{E})] = e^{\mu t} E[e^{\sigma^2 t^2 \mathcal{E}}] = e^{\mu t} M_{\mathcal{E}}(\sigma^2 t^2) = \frac{e^{\mu t}}{1-\sigma^2 t^2} \end{aligned}$$

For  $\mu + \sigma S\mathcal{E}$  we get

$$\begin{aligned} M_{\mu+\sigma S\mathcal{E}} &= E[e^{t(\mu+\sigma S\mathcal{E})}] = e^{\mu t} E[e^{t\sigma S\mathcal{E}}] = e^{\mu t} E[E[e^{t\sigma S\mathcal{E}} | S]] \\ &= e^{\mu t} E[M_{\mathcal{E}}(t\sigma S)] = e^{\mu t} E[\frac{1}{1-t\sigma S}] = e^{\mu t} (\frac{1}{2}(\frac{1}{1+t\sigma}) + \frac{1}{2}(\frac{1}{1-t\sigma})) = \frac{e^{\mu t}}{1-\sigma^2 t^2} \end{aligned}$$

Thus,  $M_{\mu+\sigma S\mathcal{E}} = M_{\mu+\sigma\sqrt{2}\mathcal{E}Z}$ , and so  $\mathcal{L} \sim \mu + \sigma\sqrt{2}\mathcal{E}Z$  (Ding and Blitzstein, 2018).

## Application to Highway Speed Data

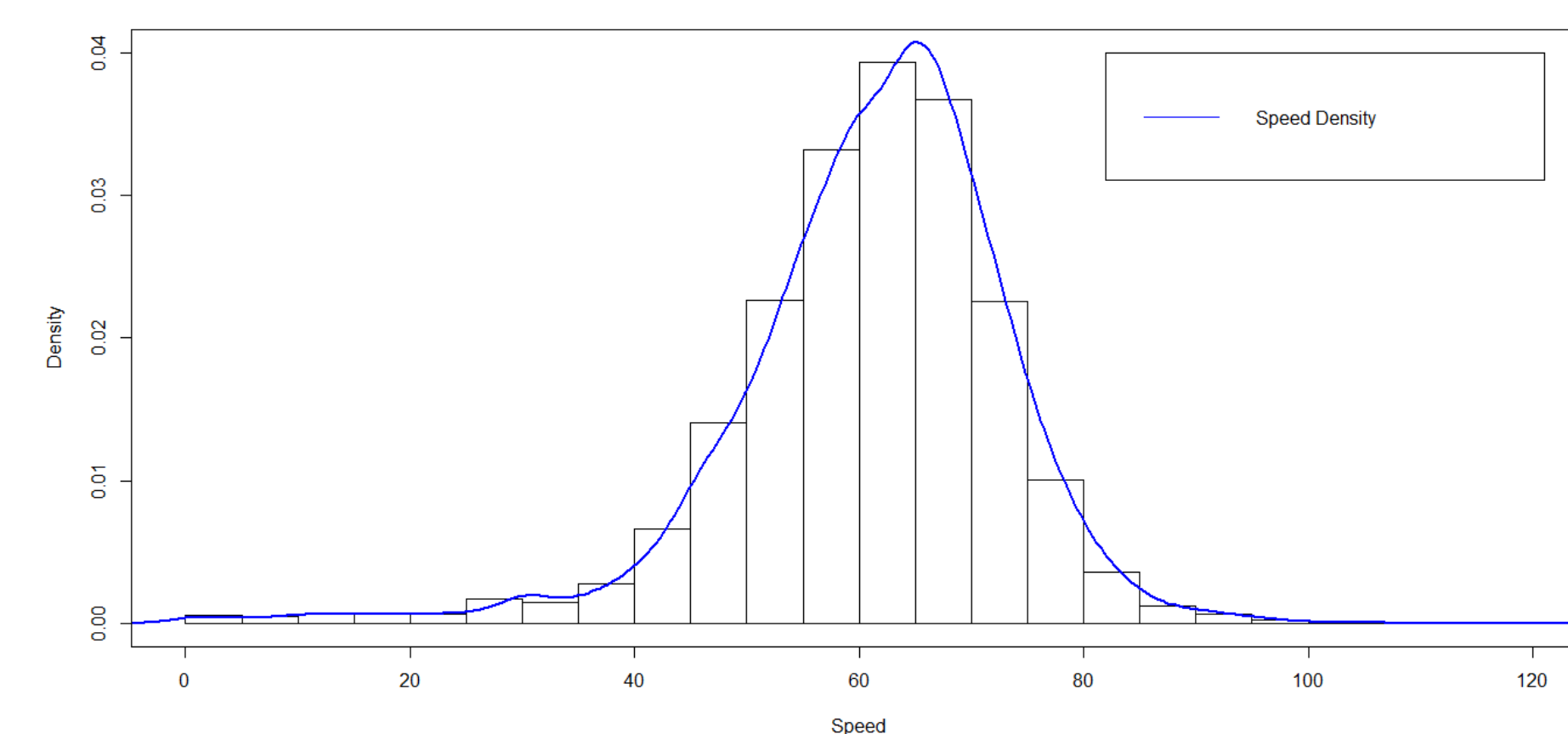


Figure 4: Highway Speed Data with Density Curve.

- Above is highway speed data taken by TDOT sensors.
- The speed data has two sources of variation: from the speed and from the measuring device. Then the Laplace distribution may be a better model than the Normal distribution (Geraci and Borja, 2018).
- Peaks relatively sharply, but possible skew is visible.
- This introduces the necessity to consider an asymmetric Laplace distribution in addition to our symmetric one.

## An Asymmetric Laplace Distribution

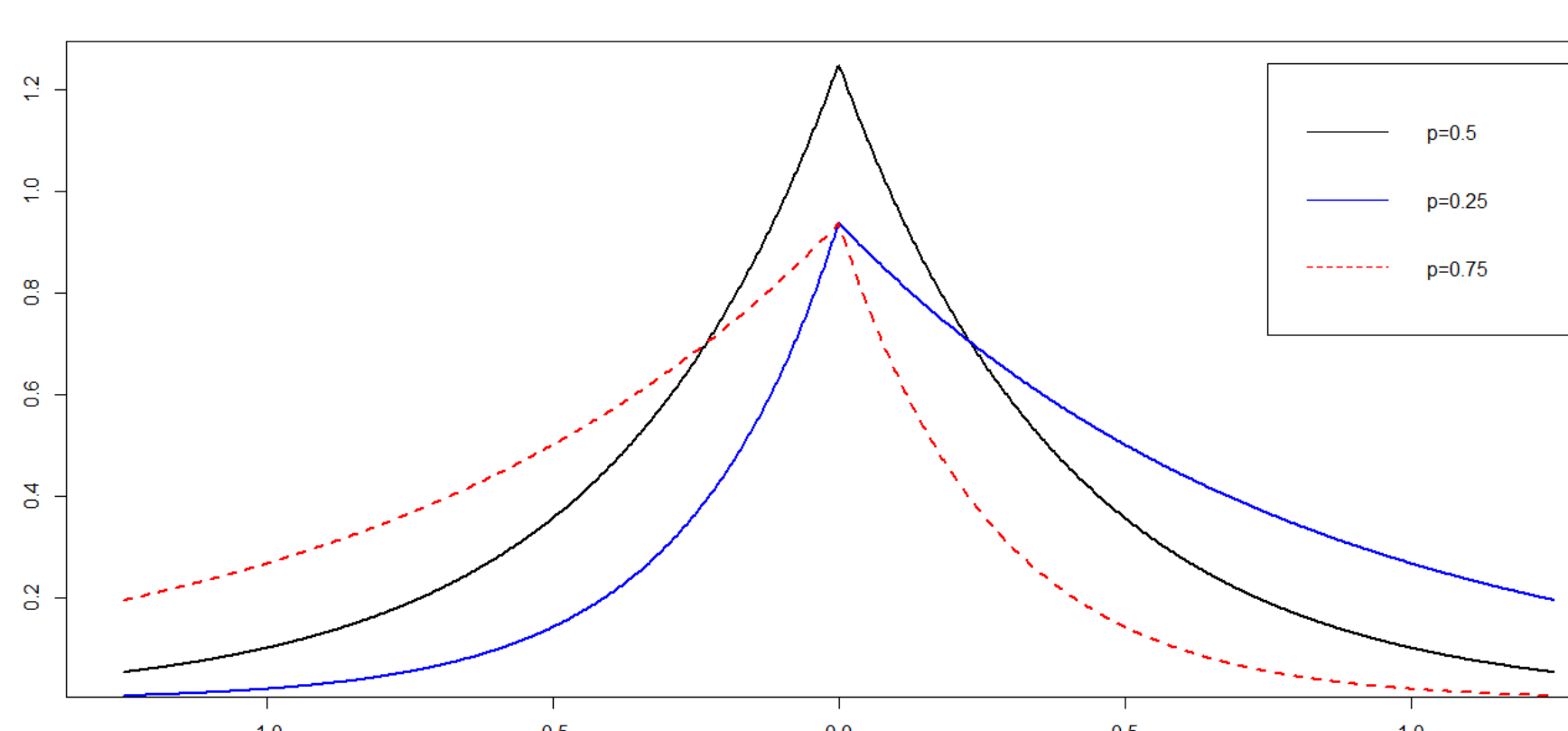


Figure 5: Effects of p on Asymmetric Laplace Distribution.

- Laplace distribution with a built in skewness parameter p which is between 0 and 1.
- Has CDF (Yu and Zhang, 2005)

$$F(x) = \begin{cases} pe^{(1-p)(x-\mu)/\sigma} & x \leq \mu \\ 1 - (1-p)e^{-p(x-\mu)/\sigma} & x > \mu \end{cases}$$

- Symmetric Laplace distribution when p=0.5.
- Using the R packages **ald**, we can fit our data.

## Modelling Results

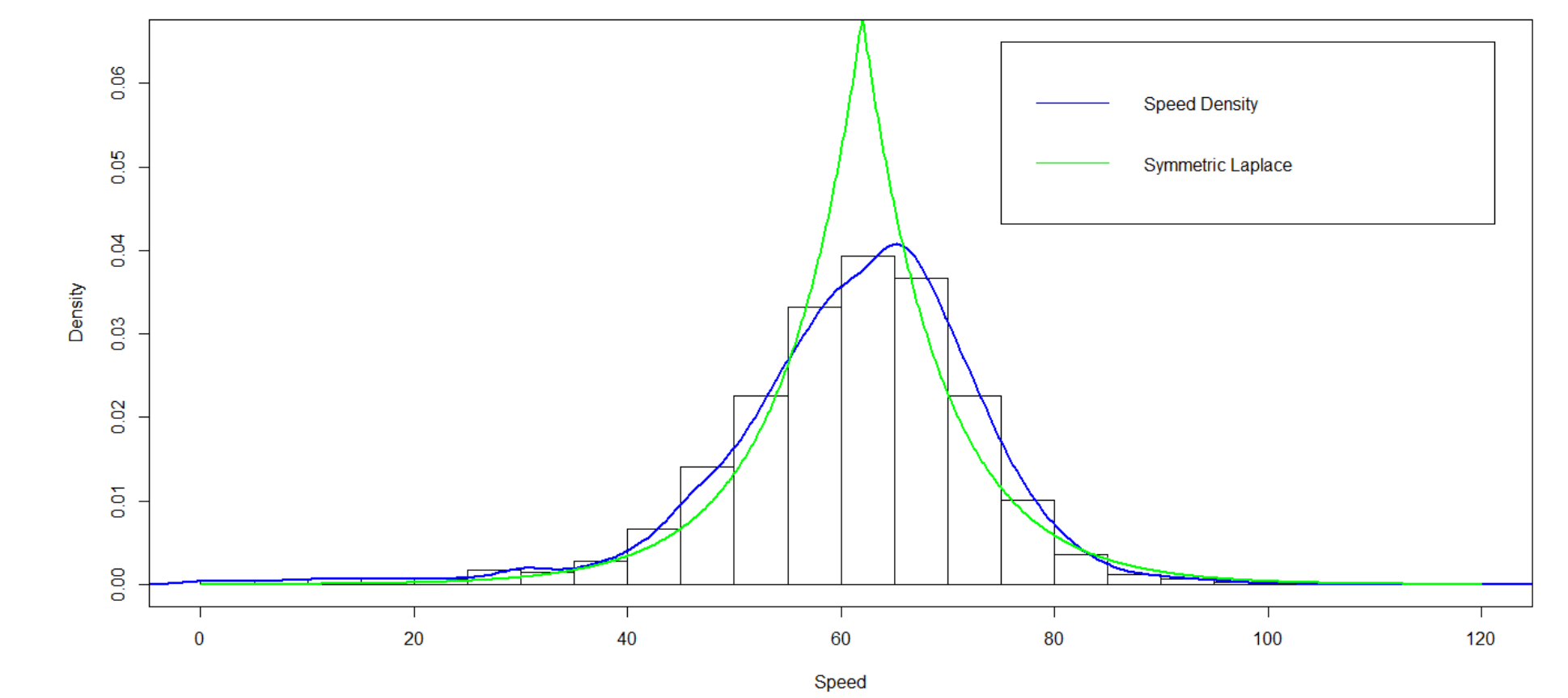


Figure 6: Laplace Distribution on Speed Data.

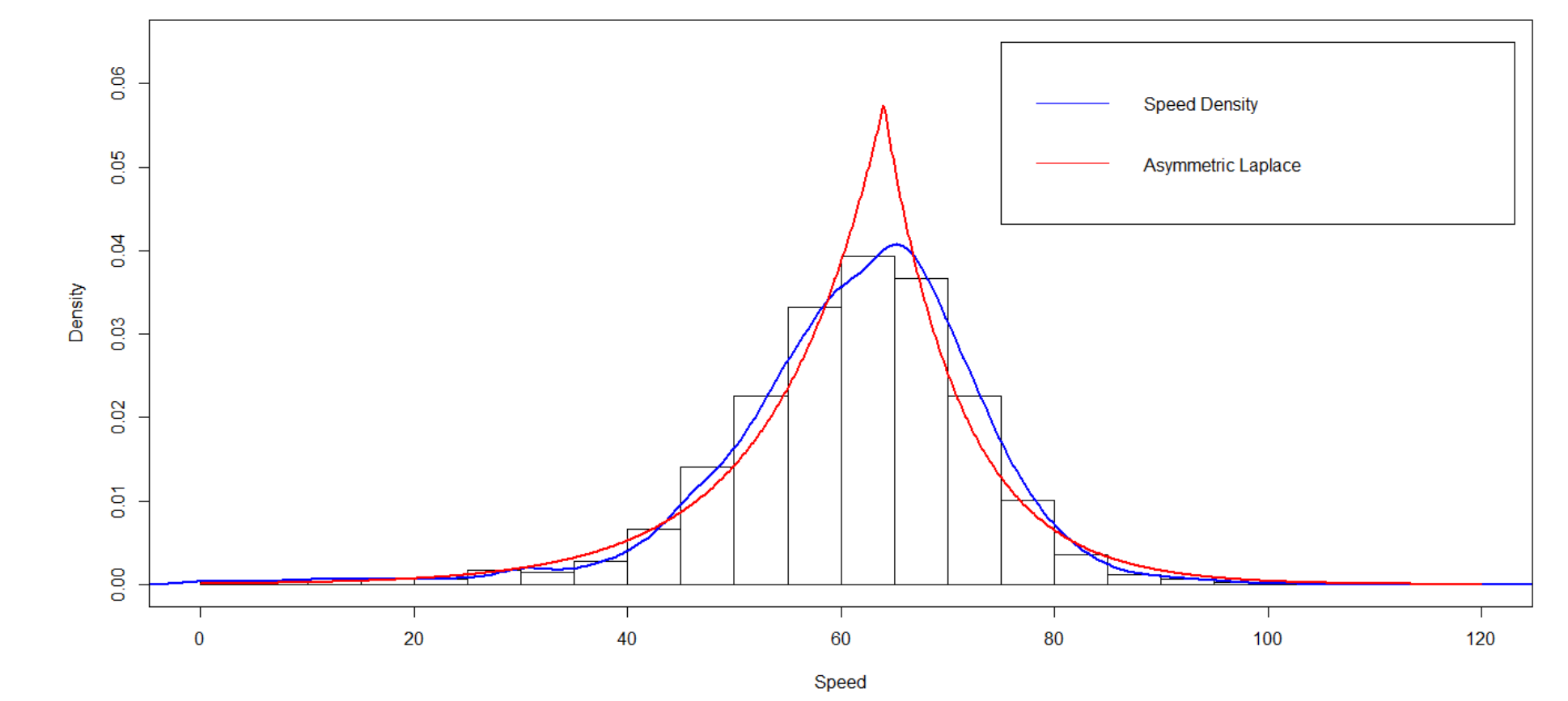


Figure 7: Asymmetric Laplace Distribution on Speed Data.

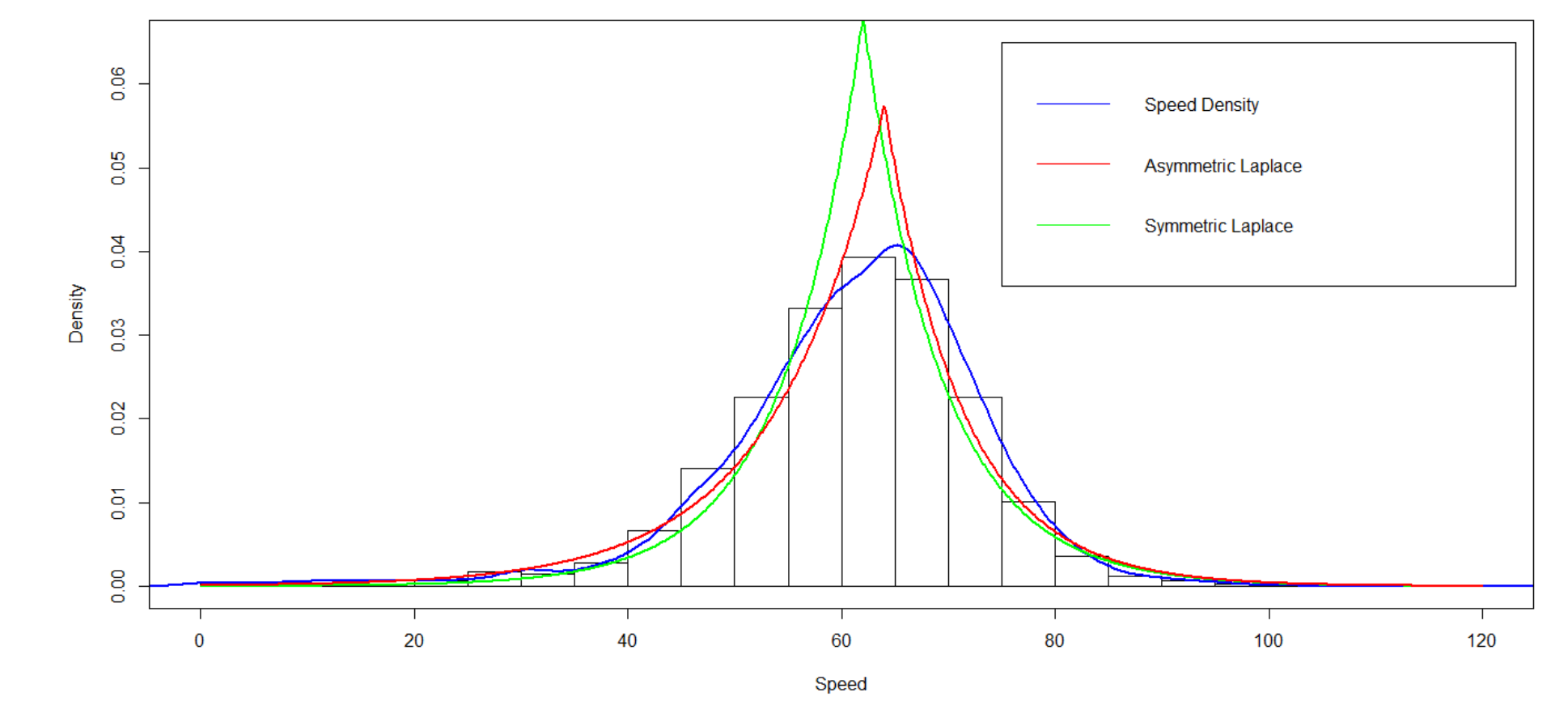


Figure 8: Laplace and Asymmetric Laplace Distributions on Speed Data.

- Using built in MLE functions in **ald**, we first estimate parameters for the symmetric and asymmetric Laplace distributions (Galarza and Lachos, 2018).
- By inspection, we see that the asymmetric Laplace distribution seems to fit better.
- To verify this we use Akaike information criterion:
  - For the symmetric distribution, we get  $AIC_1 \approx 6.0 \times 10^6$ .
  - For the asymmetric distribution, we get  $AIC_2 \approx 5.6 \times 10^6$ .
- We get  $AIC_1 > AIC_2$ , and so the asymmetric model is verified to be a better fit.

## Selected References

- Ding, Peng and Blitzstein, Joseph K. (2018), On the Gaussian Mixture Representation of the Laplace Distribution, *The American Statistician*, 72:2, 172-174, DOI: 10.1080/00031305.2017.1291448.
- Galarza, Christian E. and Lachos, Victor H. (2018), The Asymmetric Laplace Distribution, R package version 1.2.
- Geraci, M. and Borja, M. C. (2018), Notebook: The Laplace distribution. *Significance*, 15: 10-11. DOI:10.1111/j.1740-9713.2018.01185.x.
- Kotz, Samuel and Kozubowski, Tomasz and Podgorski, Krzysztof (2001), The Laplace Distribution and Generalizations.
- Yu, Keming and Zhang, Jin (2005), A Three-Parameter Asymmetric Laplace Distribution and Its Extension, *Communications in Statistics - Theory and Methods*, 34:9-10, 1867-1879, DOI: 10.1080/03610920500199018.