

# The Weibull and Gumbel (Extreme Value) Distributions<sup>1</sup>

STA312 Fall 2023

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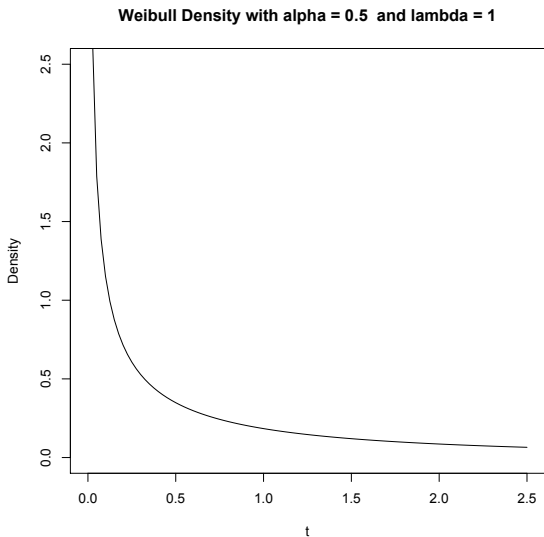
<sup>1</sup>See last slide for copyright information.

## The Weibull Distribution

$$f(t|\alpha, \lambda) = \begin{cases} \alpha\lambda(\lambda t)^{\alpha-1} \exp\{-(\lambda t)^\alpha\} & \text{for } t \geq 0 \\ 0 & \text{for } t < 0 \end{cases}$$

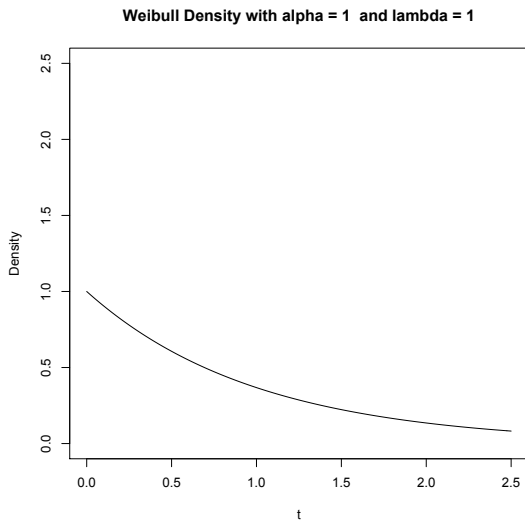
where  $\alpha > 0$  and  $\lambda > 0$ .

# Weibull with $\alpha = 1/2$ and $\lambda = 1$

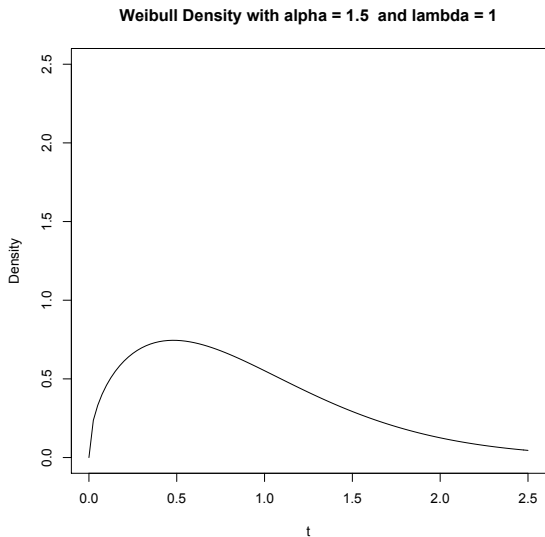


# Weibull with $\alpha = 1$ and $\lambda = 1$

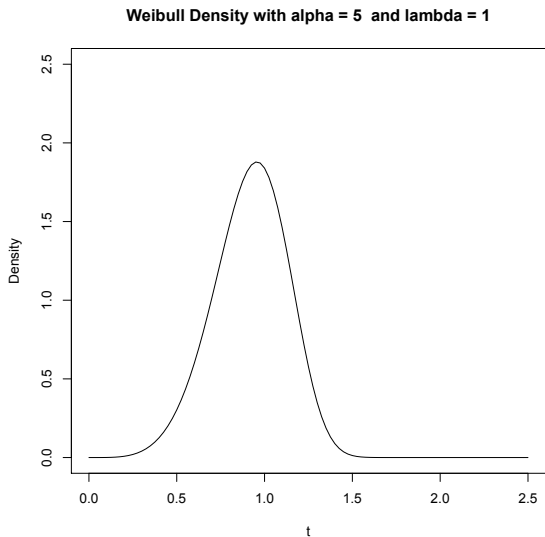
Standard exponential



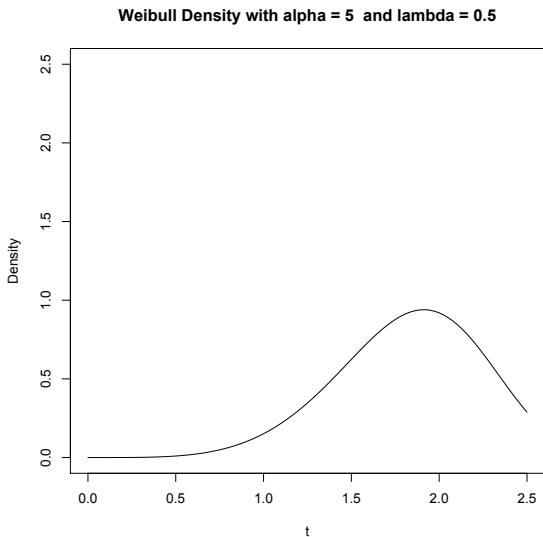
# Weibull with $\alpha = 1.5$ and $\lambda = 1$



# Weibull with $\alpha = 5$ and $\lambda = 1$



# Weibull with $\alpha = 5$ and $\lambda = 1/2$



# The Weibull Distribution

$$f(t|\alpha, \lambda) = \begin{cases} \alpha\lambda(\lambda t)^{\alpha-1} \exp\{-(\lambda t)^\alpha\} & \text{for } t \geq 0 \\ 0 & \text{for } t < 0 \end{cases},$$

where  $\alpha > 0$  and  $\lambda > 0$ .

$$\begin{aligned} E(T^k) &= \frac{\Gamma(1 + \frac{k}{\alpha})}{\lambda^k} \\ \text{Median} &= \frac{[\log(2)]^{1/\alpha}}{\lambda} \\ S(t) &= \exp\{-(\lambda t)^\alpha\} \\ h(t) &= \alpha\lambda^\alpha t^{\alpha-1} \end{aligned}$$

- If  $\alpha = 1$ , Weibull reduces to exponential and  $h(t) = \lambda$ .
- If  $\alpha > 1$ , the hazard function is increasing in  $t$ .
- If  $\alpha < 1$ , the hazard function is decreasing.



# The Gumbel (or Extreme Value) Distribution

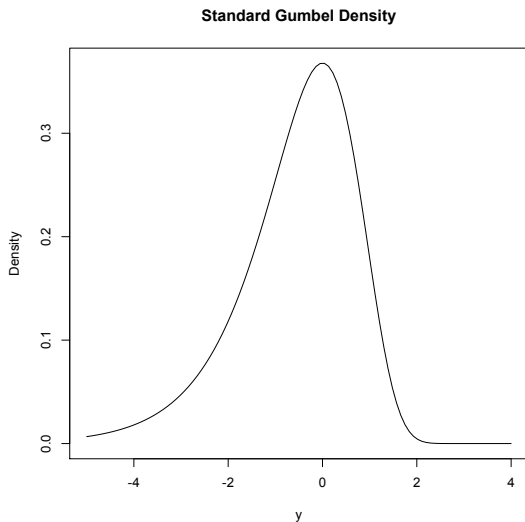
This version is based on the log of an exponential, not  $-\log$  as in HW4

$$f(y|\mu, \sigma) = \frac{1}{\sigma} \exp \left\{ \left( \frac{y - \mu}{\sigma} \right) - e^{\left( \frac{y - \mu}{\sigma} \right)} \right\}$$

where  $\sigma > 0$ .

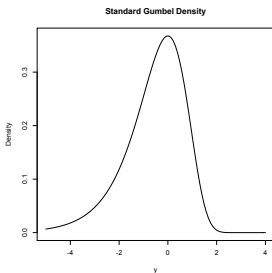
- This is a location-scale family of distributions.
- $\mu$  is the location and  $\sigma$  is the scale.
- Write  $Y \sim G(\mu, \sigma)$ .

Log (not  $-\log$ ) of standard exponential is Gumbel(0,1)  
 $\mu = 0$  and  $\sigma = 1$



# Properties of the $G(0, 1)$ Distribution

$f(y) = \exp\{y - e^y\}$  for all real  $y$ .



Let  $Z \sim G(0, 1)$ .

- MGF is  $M_z(t) = \Gamma(t + 1)$ .
- $E(Z) = \Gamma'(1) = -0.5772157\dots = -\gamma$ , where  $\gamma$  is Euler's constant.
- $Var(Z) = \frac{\pi^2}{6}$ .
- Median is  $\log(\log(2)) = -0.3665129\dots$
- Mode is zero.

General  $Y \sim G(\mu, \sigma)$

$$f(y|\mu, \sigma) = \frac{1}{\sigma} \exp \left\{ \left( \frac{y-\mu}{\sigma} \right) - e^{\left( \frac{y-\mu}{\sigma} \right)} \right\}$$

Let  $Z \sim G(0, 1)$  and  $Y = \sigma Z + \mu$ . Then  $Y \sim G(\mu, \sigma)$ .

- $E(Y) = \sigma E(Z) + \mu = \mu - \sigma\gamma$ .
- $Var(Y) = \sigma^2 Var(Z) = \sigma^2 \frac{\pi^2}{6}$ .
- Median is  $\sigma \log \log(2) + \mu$ .
- Mode is  $\mu$ .

## Log (not minus log) of Weibull is Gumbel

- Let  $T \sim \text{Weibull}(\alpha, \lambda)$ , and  $Y = \log(T)$ .
- In addition, re-parameterize, meaning express the parameters in a different, equivalent way.
- Let  $\sigma = \frac{1}{\alpha}$  and  $\mu = -\log \lambda$ .
- Or equivalently, substitute  $\frac{1}{\sigma}$  for  $\alpha$  and  $e^{-\mu}$  for  $\lambda$ .
- The result is  $Y \sim G(\mu, \sigma)$ .
  
- So if you believe the distribution of a set of failure time data could be Weibull (a popular choice), you can log-transform the data and apply a Gumbel model.
- The Gumbel distribution may be preferable because the parameters  $\mu$  and  $\sigma$  are easy to interpret.

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<http://www.utstat.toronto.edu/brunner/oldclass/312f23>