



UNCERTAINTY AND RELIABILITY



OUTLINE

- Sources of Uncertainty
- Background information
 1. Probability basics
 2. Reliability/Performance Models
- Different sources of information



SOURCES OF UNCERTAINTY

- “Statistical” Variability
 1. sampling variability
 2. measurement error

- Data

- Bias

- Model



SOURCES OF UNCERTAINTY

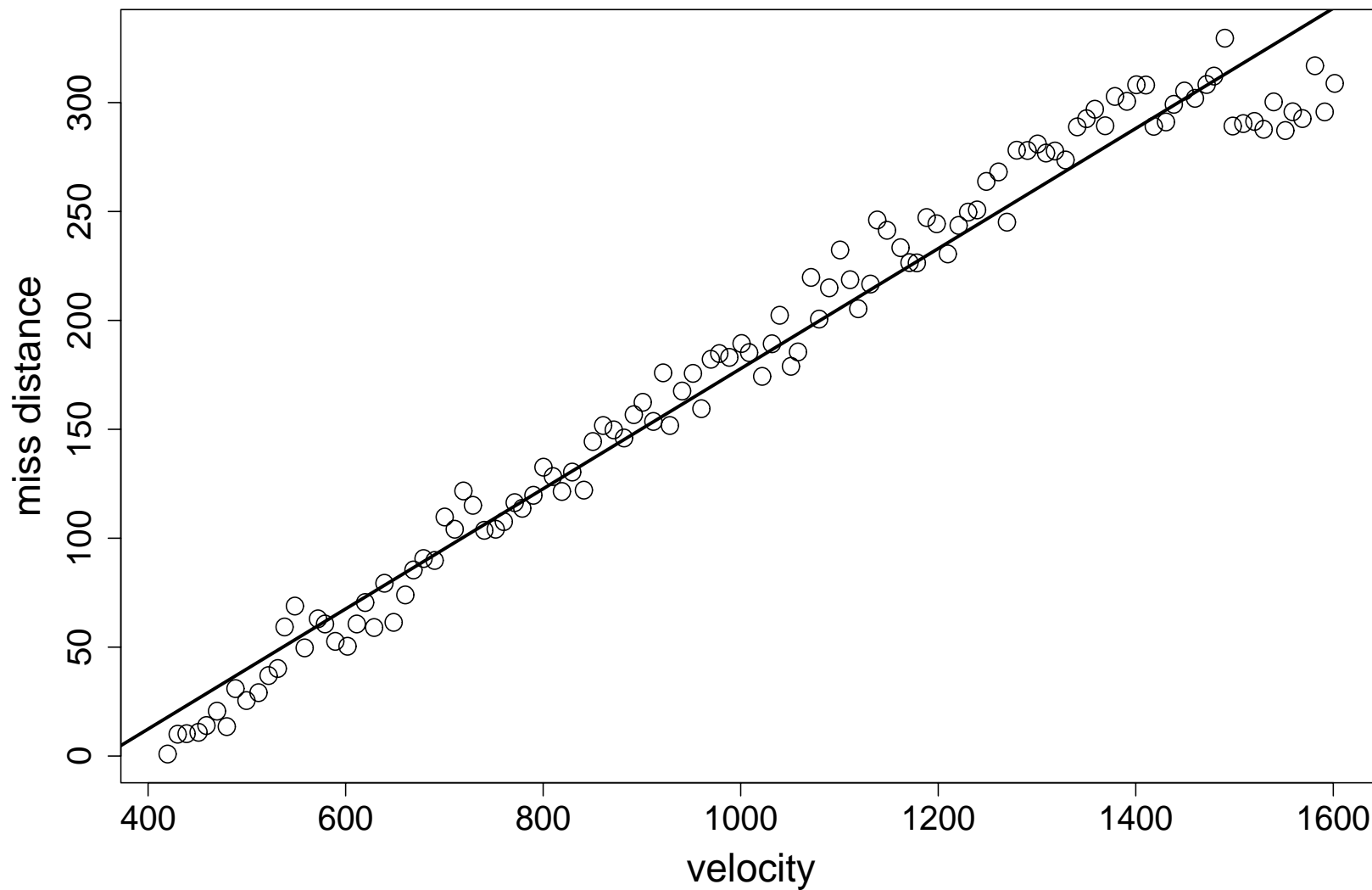
Example

- Want to understand the effect of threat characteristics on missile performance
- Response variable: miss distance
- Explanatory variable: velocity of the threat



SOURCES OF UNCERTAINTY

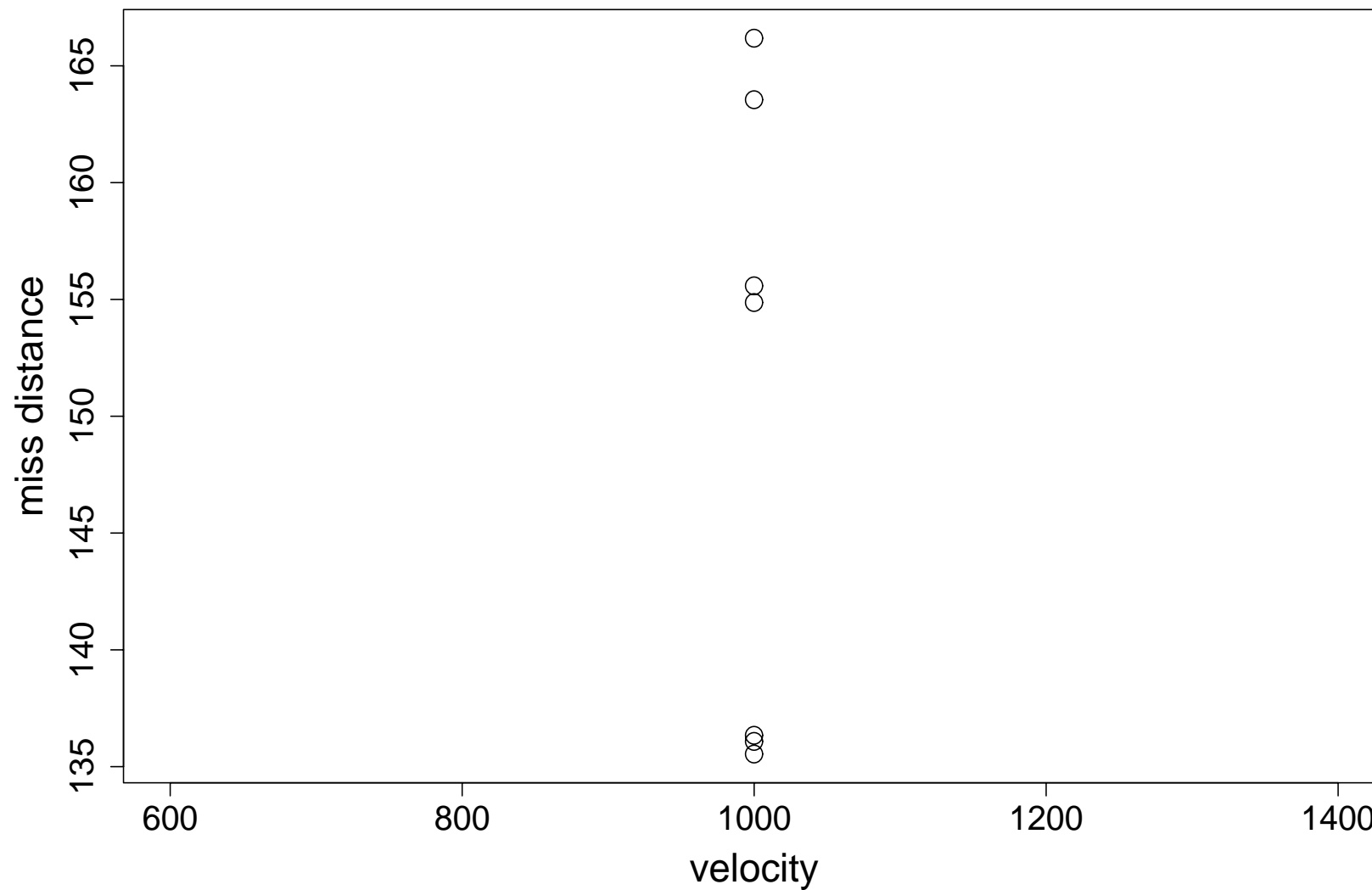
Threat Characteristic Assessment





SOURCES OF UNCERTAINTY

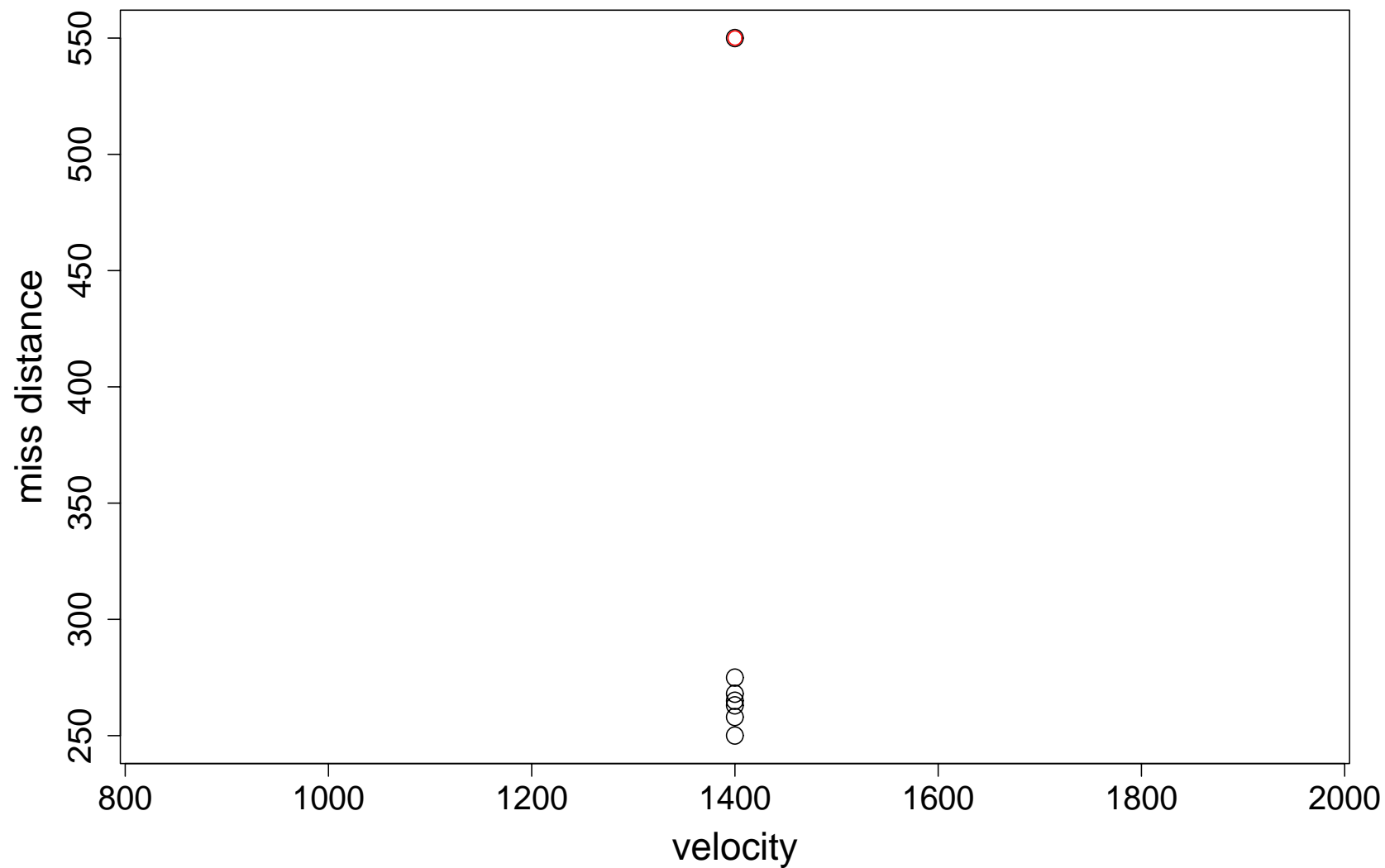
Sampling Variability





SOURCES OF UNCERTAINTY

Data Uncertainty





SOURCES OF UNCERTAINTY

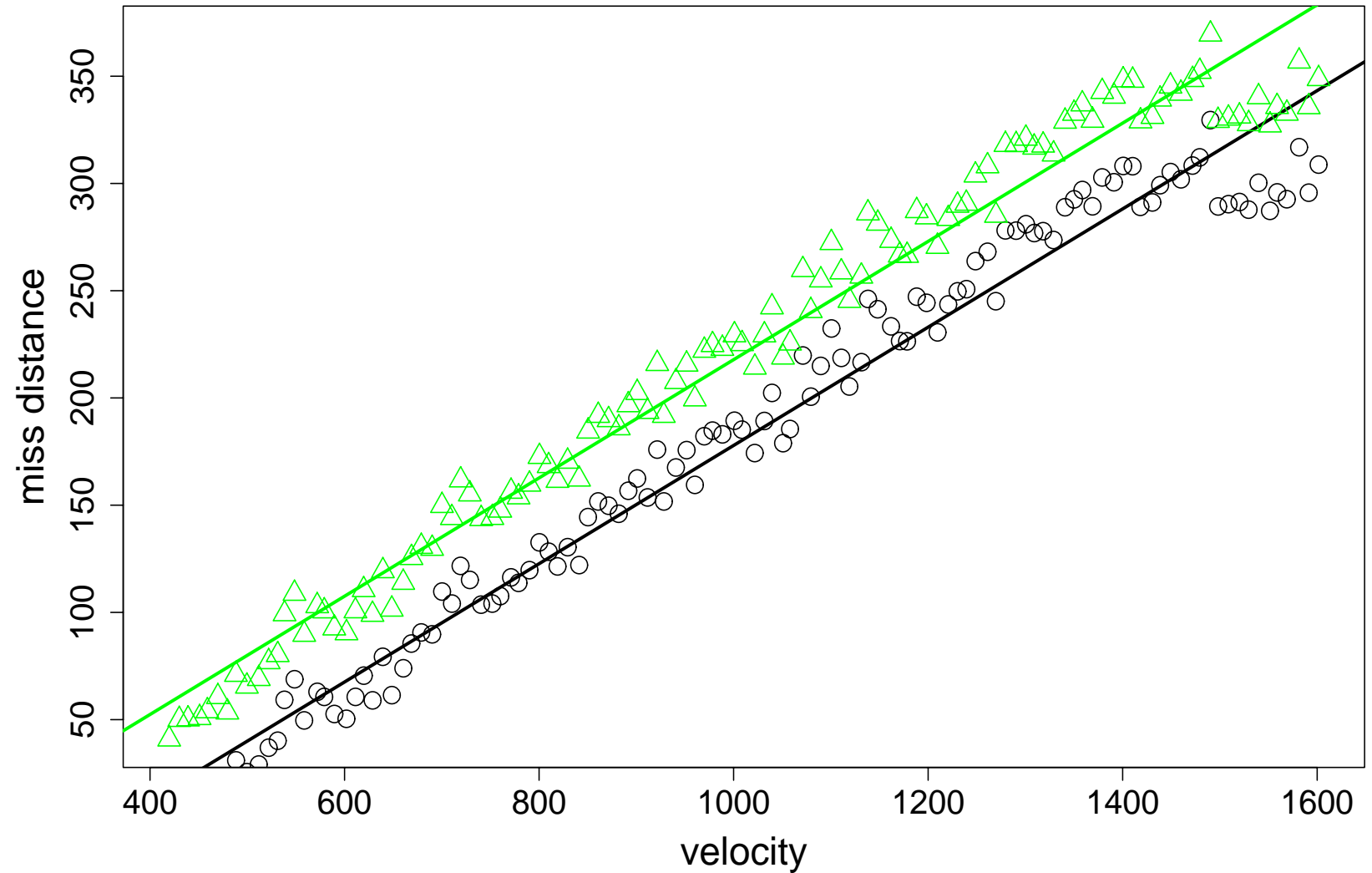
Computer Models May Induce Bias

- Physical experiments: everything's in there, we just don't know what's in there
- Computer experiments: **not** everything's in there, but we know everything that's in there.
- two kinds of bias:
 1. location bias
 2. scale bias



SOURCES OF UNCERTAINTY

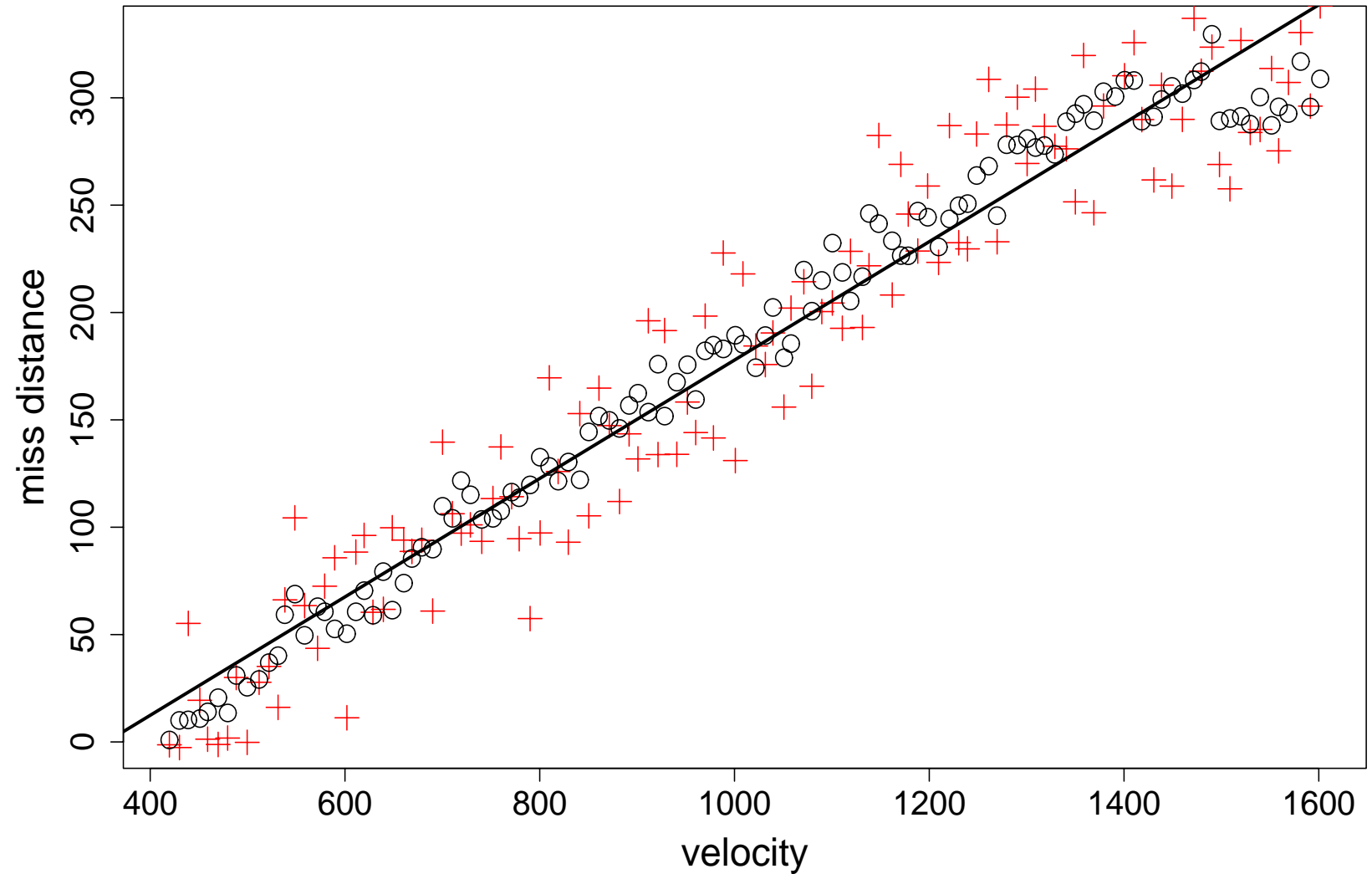
Location Bias





SOURCES OF UNCERTAINTY

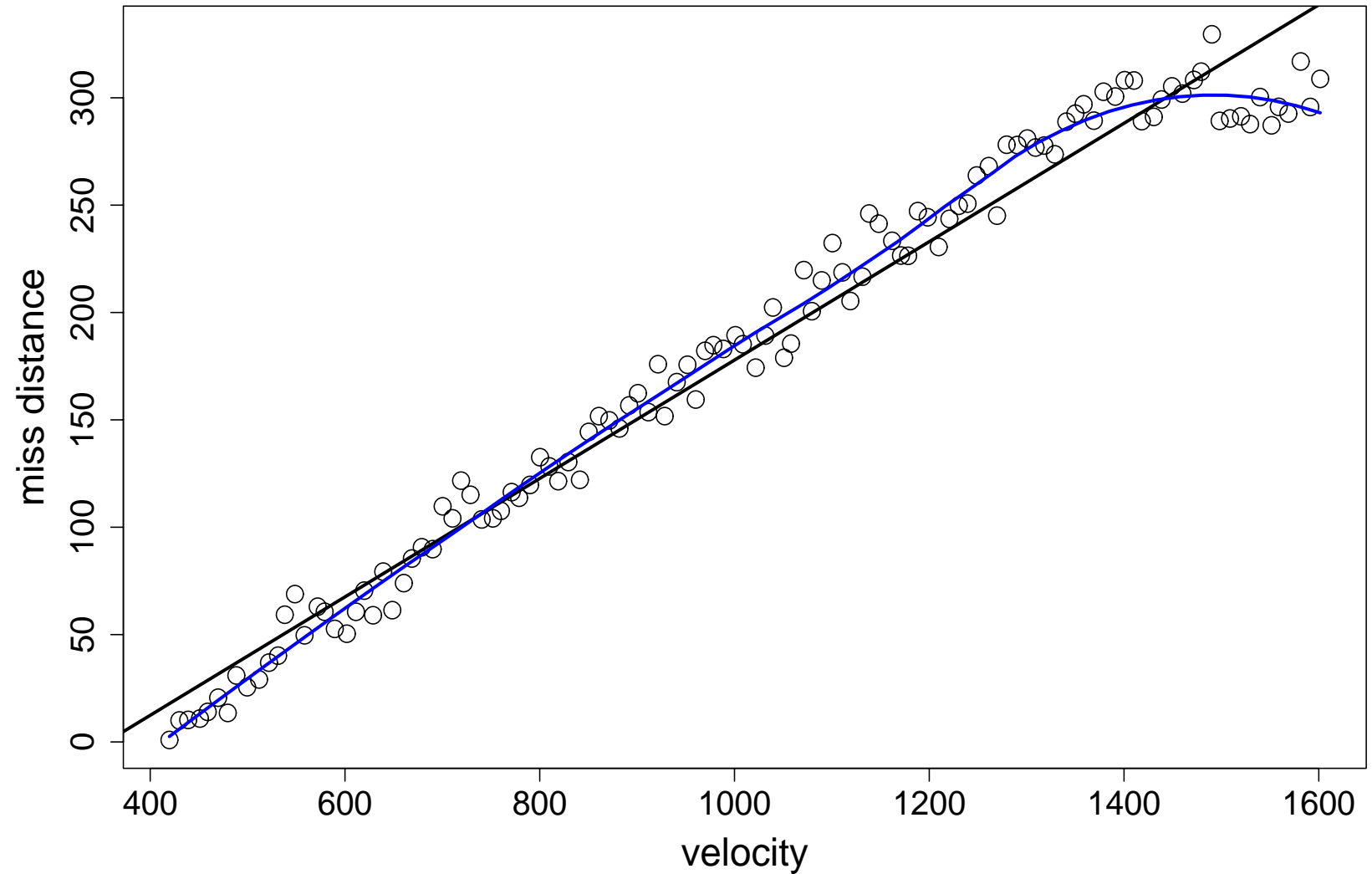
Scale Bias





SOURCES OF UNCERTAINTY

Model Uncertainty





DISCUSSION

- Difficulties
 1. Statistical – need data (sometimes *a lot*)
 2. Data – exactly what constitutes “strange”?
 3. Bias – identifiability
 4. Model – by far the hardest to assess
- WIP



CONDITIONAL PROBABILITY

- Independence: $Pr(A \cap B) = Pr(A) \times Pr(B)$
- Conditional Probability

$$Pr(A|B) = \frac{Pr(A \cap B)}{Pr(B)}$$

Note: if A and B are independent, then
 $Pr(A|B) = Pr(A)$.



RELIABILITY MODELS

- What is reliability?

$$R(t) = Pr(T \geq t) = \int_0^t f(x)dx$$

where $f(x)$ is the distribution of failure times.

- Easy textbook definition, hard in practice.
 1. What is a failure?
 2. Is there a mapping between performance and reliability?



RELIABILITY MODELS

slide indicating problems defining reliability
reliability vs. performance



RELIABILITY MODELS

Hazard Rate

- Define *hazard rate* as:

$$h(t) = \frac{f(t)}{R(t)}$$

- So what?
 1. Instantaneous failure rate
 2. Great for model identification



RELIABILITY

Reliability Basics

- T represents our *random variable* of interest.
- Often T is the time until failure or failure time.
- Probability distribution of failure times ($f(t)$).
- t represents the *realization* of the random variable.



RELIABILITY – BINOMIAL

- Model for success/failure data
- n trials, X successes
- $p = Pr(\text{success})$ is the same for all trials
- Trials are independent
-

$$Pr(X = c) = \binom{n}{c} p^c (1 - p)^{n-c}$$



RELIABILITY – EXPONENTIAL

- Reliability

$$R(t|\lambda) = \exp^{-\lambda t}, t > 0$$

- Hazard Rate

$$h(t|\lambda) = \lambda$$

- Mean Time to Failure (MTTF)

$$E(T) = \frac{1}{\lambda}$$

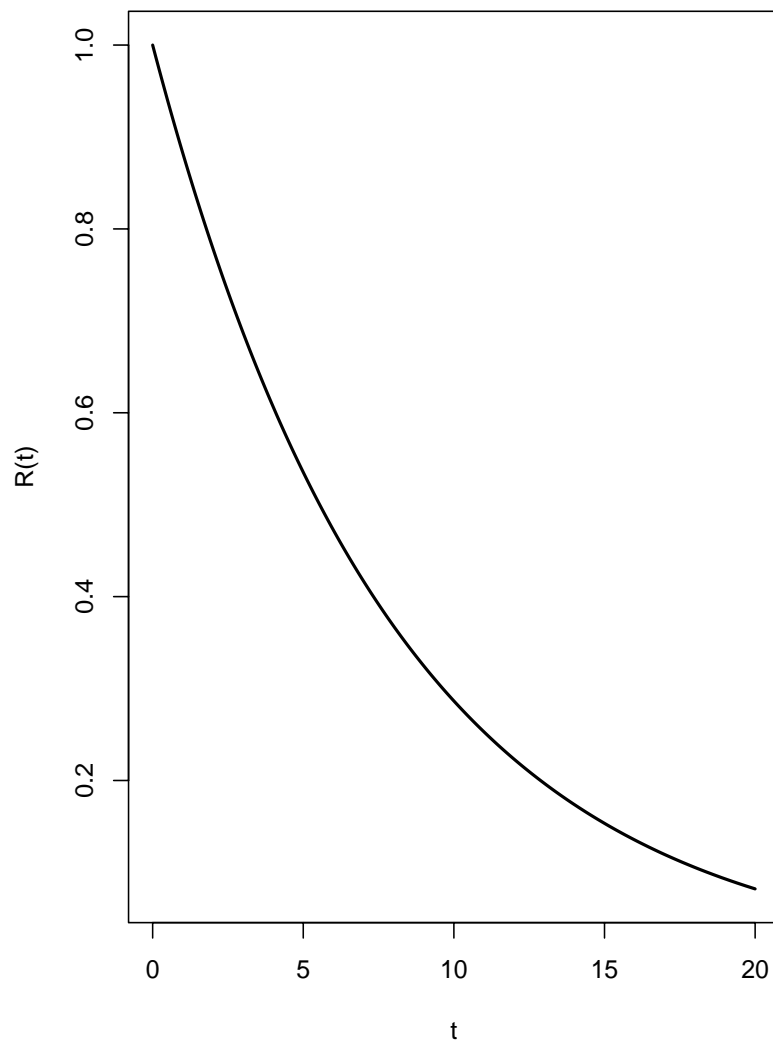
- Variance of TTF

$$V(T) = \frac{1}{\lambda^2}$$

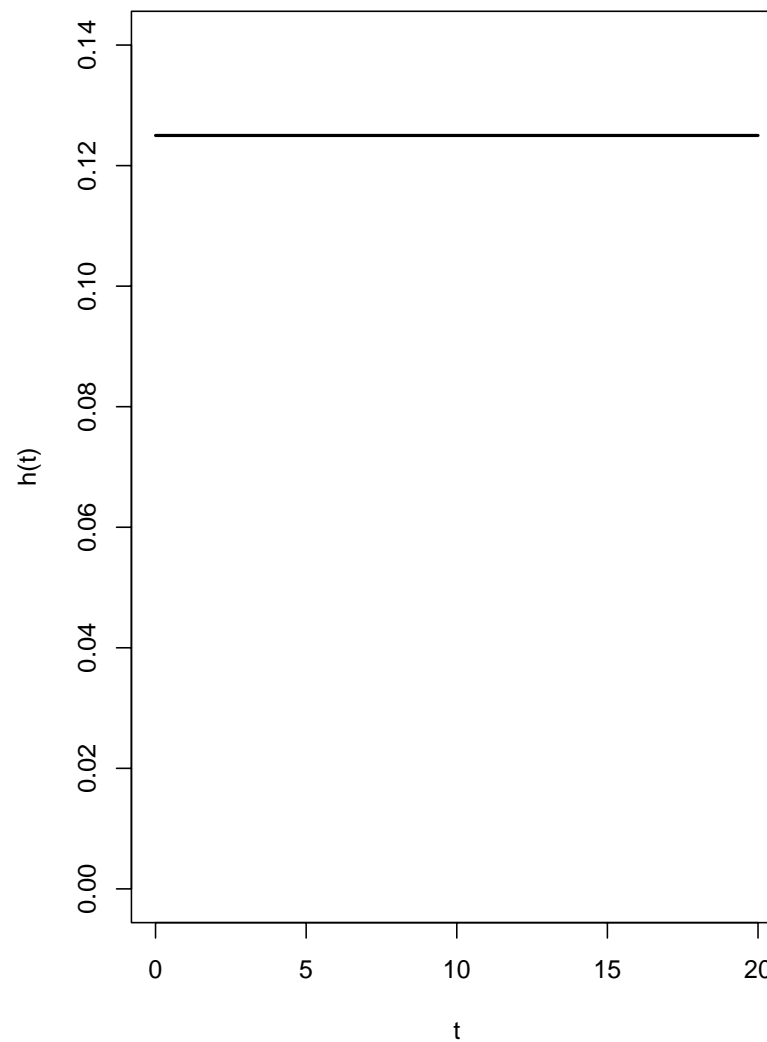


RELIABILITY – EXPONENTIAL

Reliability



Hazard





RELIABILITY – WEIBULL

- Reliability

$$R(t|\alpha, \theta) = \exp \left[- \left(\frac{t - \theta}{\alpha} \right) \right], t > \theta$$

- Hazard Rate

$$h(t|\alpha, \theta) = \frac{\beta}{\alpha} \left(\frac{t - \theta}{\alpha} \right)^{\beta-1}, t > \theta$$

- Mean Time to Failure (MTTF)

$$E(T) = \theta + \alpha \Gamma \left(\frac{\beta + 1}{\beta} \right)$$

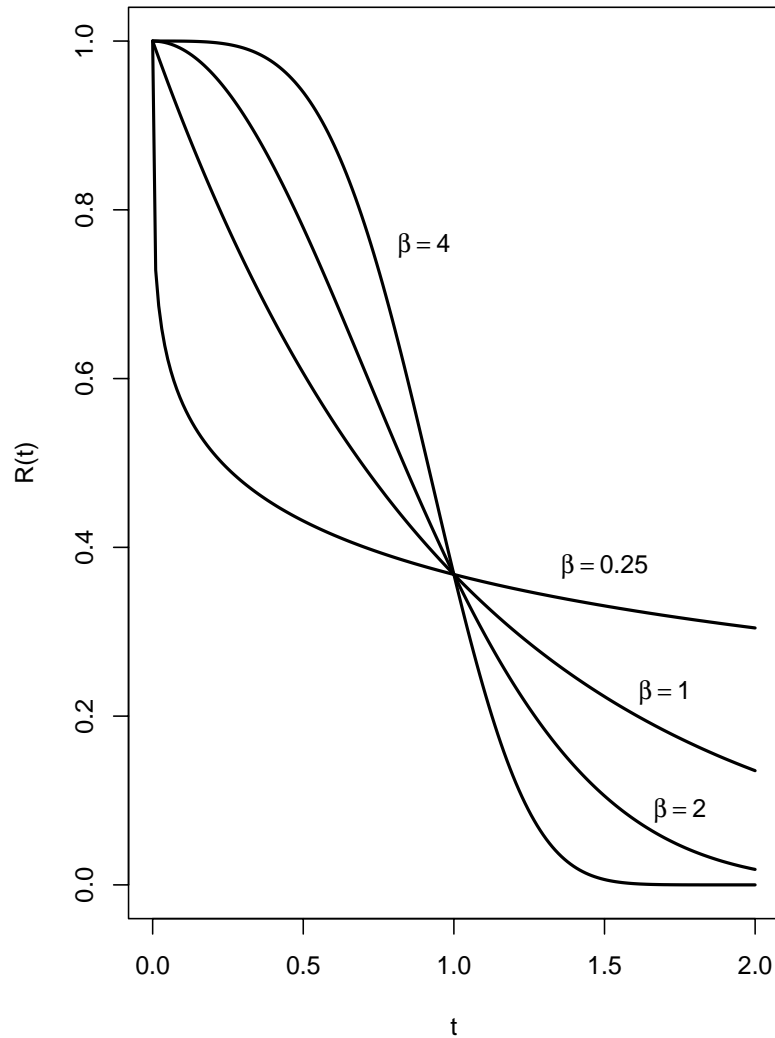
- Variance of TTF

$$V(T) = \alpha^2 \left[\Gamma \left(\frac{\beta + 2}{\beta} \right) - \Gamma^2 \left(\frac{\beta + 1}{\beta} \right) \right]$$

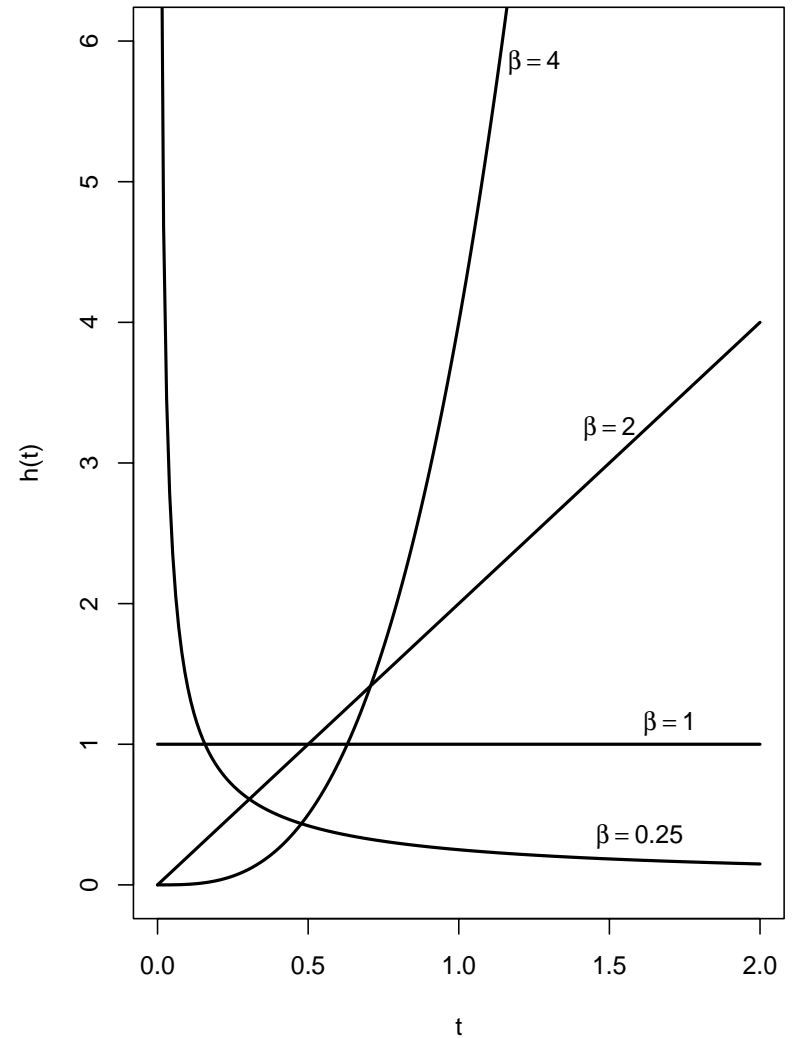


RELIABILITY – WEIBULL

Reliability



Hazard





RELIABILITY – LOGNORMAL

- Reliability: if $\log(Y) \sim Normal$ then $Y \sim LN$

$$R(t|\xi, \sigma) = 1 - \Phi\left(\frac{\log t - \xi}{\sigma}\right)$$

- Hazard Rate

$$h(t|\xi, \sigma) = \frac{\phi\left(\frac{\log t - \xi}{\sigma}\right)}{\sigma t - \sigma t \Phi\left(\frac{\log t - \xi}{\sigma}\right)}$$

- Mean Time to Failure (MTTF)

$$E(T) = \exp\left(\xi + \frac{\sigma^2}{2}\right)$$

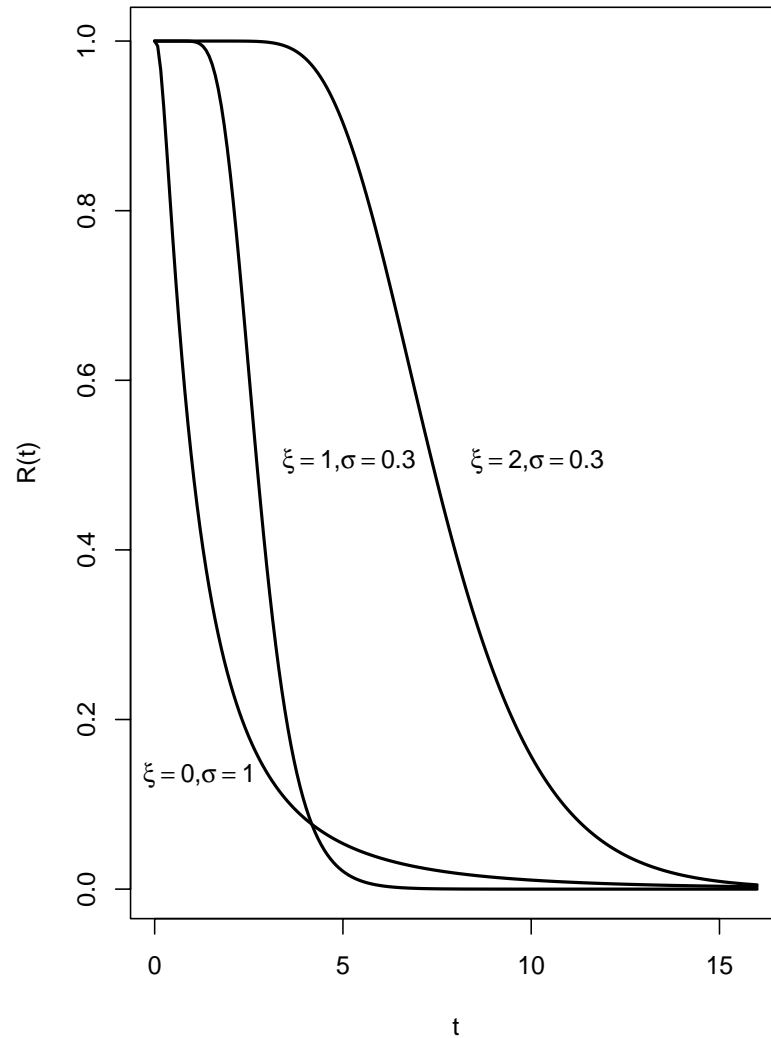
- Variance of TTF

$$V(T) = (e^{2\xi + \sigma^2})(e^{\sigma^2} - 1)$$

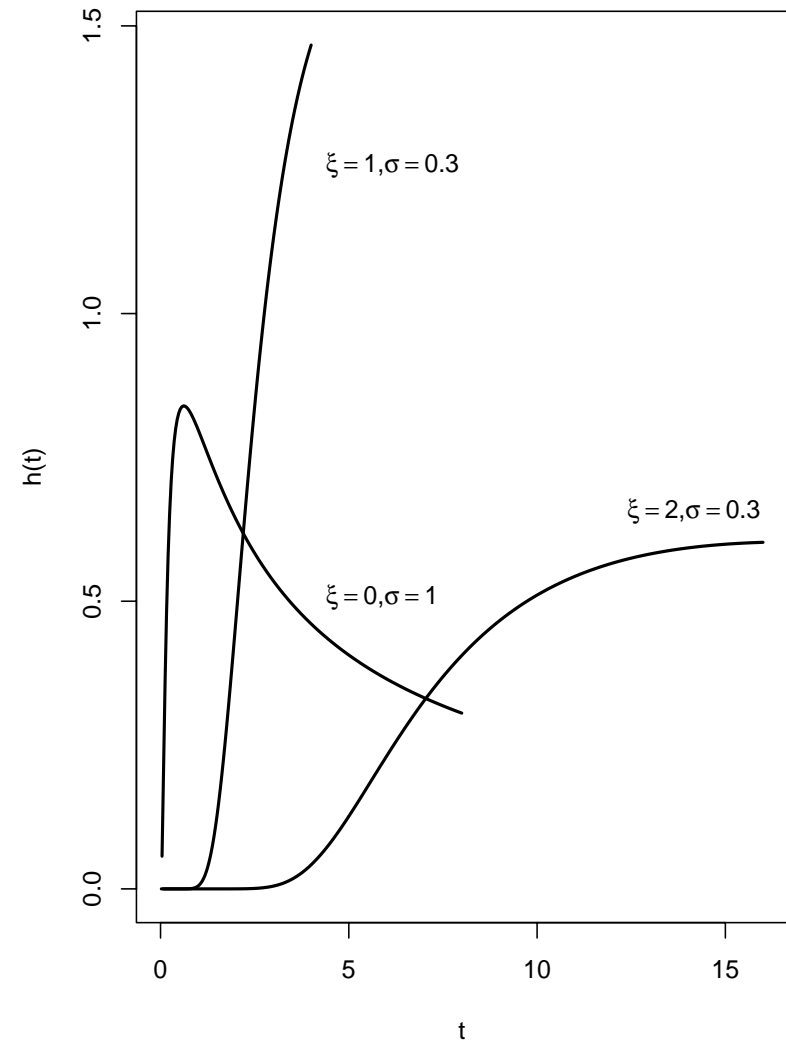


RELIABILITY – LOGNORMAL

Reliability



Hazard





RELIABILITY – GAMMA

- Reliability:

$$R(t|\alpha, \beta) = \frac{\Gamma(\alpha) - \Gamma(\alpha, t\beta)}{\Gamma(\alpha)}$$

- Hazard Rate

$$h(t|\alpha, \beta) = \beta^\alpha \frac{t^{\alpha-1} \exp(-t\beta)}{[\Gamma(\alpha) - \Gamma(\alpha, t\beta)]}$$

- Mean Time to Failure (MTTF)

$$E(T) = \frac{\alpha}{\beta}$$

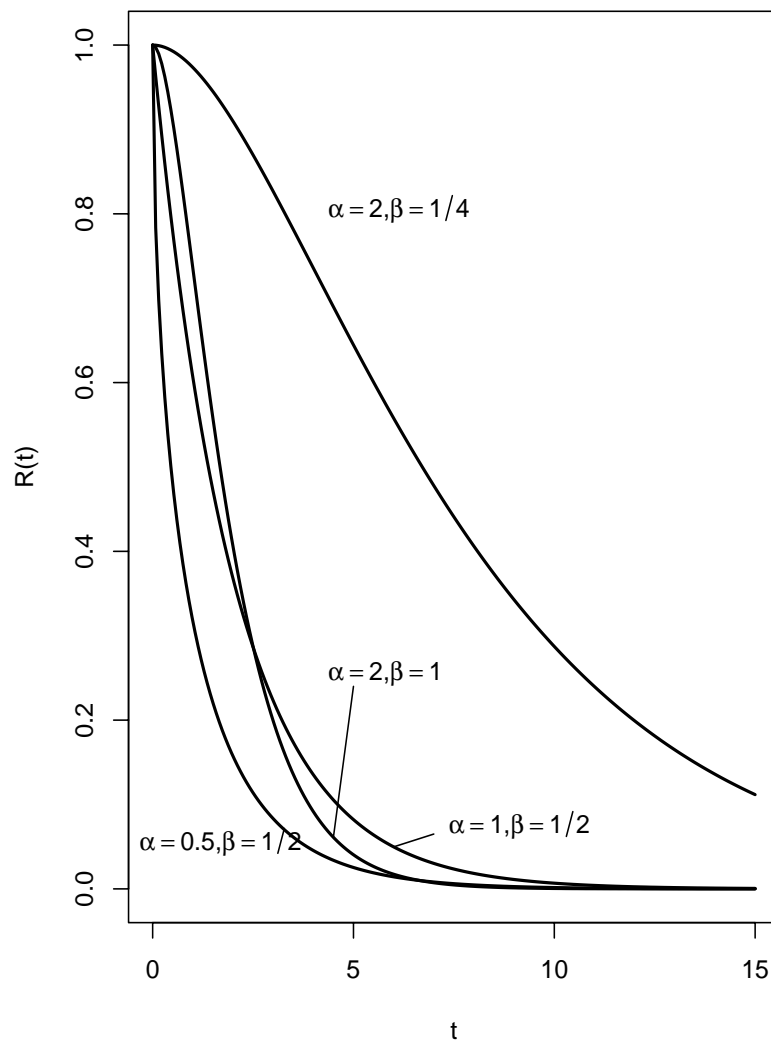
- Variance of TTF

$$V(T) = \frac{\alpha}{\beta^2}$$

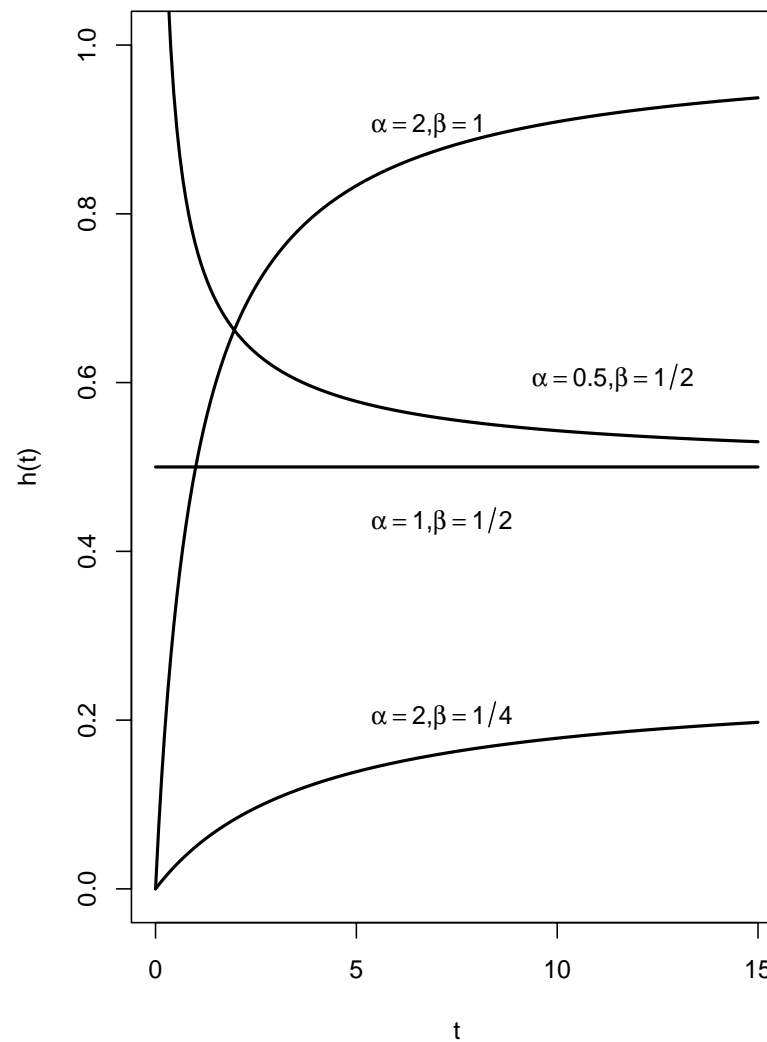


RELIABILITY – GAMMA

Reliability



Hazard





PERFORMANCE

Performance Basics

- Y represents our *random variable* of interest.
- Represents the level of performance and its uncertainty.
- Probability distribution of failure times ($g(y)$).
- y represents the *realization* of the random variable.



PERFORMANCE – NORMAL

- Performance Distribution:

$$g(y|\mu, \sigma) = (2\pi)^{-\frac{1}{2}} \exp \left\{ -\frac{(y - \mu)^2}{2\sigma^2} \right\}$$

- Mean performance

$$E(Y) = \mu$$

- Variance of performance

$$V(Y) = \sigma^2$$

- Sometimes called “Gaussian”.



PERFORMANCE – BETA

- Performance: [support on (0,1)]

$$g(y|n_o, x_o) = \frac{\Gamma(n_o)}{\Gamma(x_o)\Gamma(n_o - x_o)} y^{x_o-1} (1 - y)^{n_o - x_o - 1}$$

- Mean performance

$$E(Y) = \frac{x_o}{n_o}$$

- Variance of performance

$$V(Y) = \frac{x_o(n_o - x_o)}{n_o^2(n_o + 1)}$$

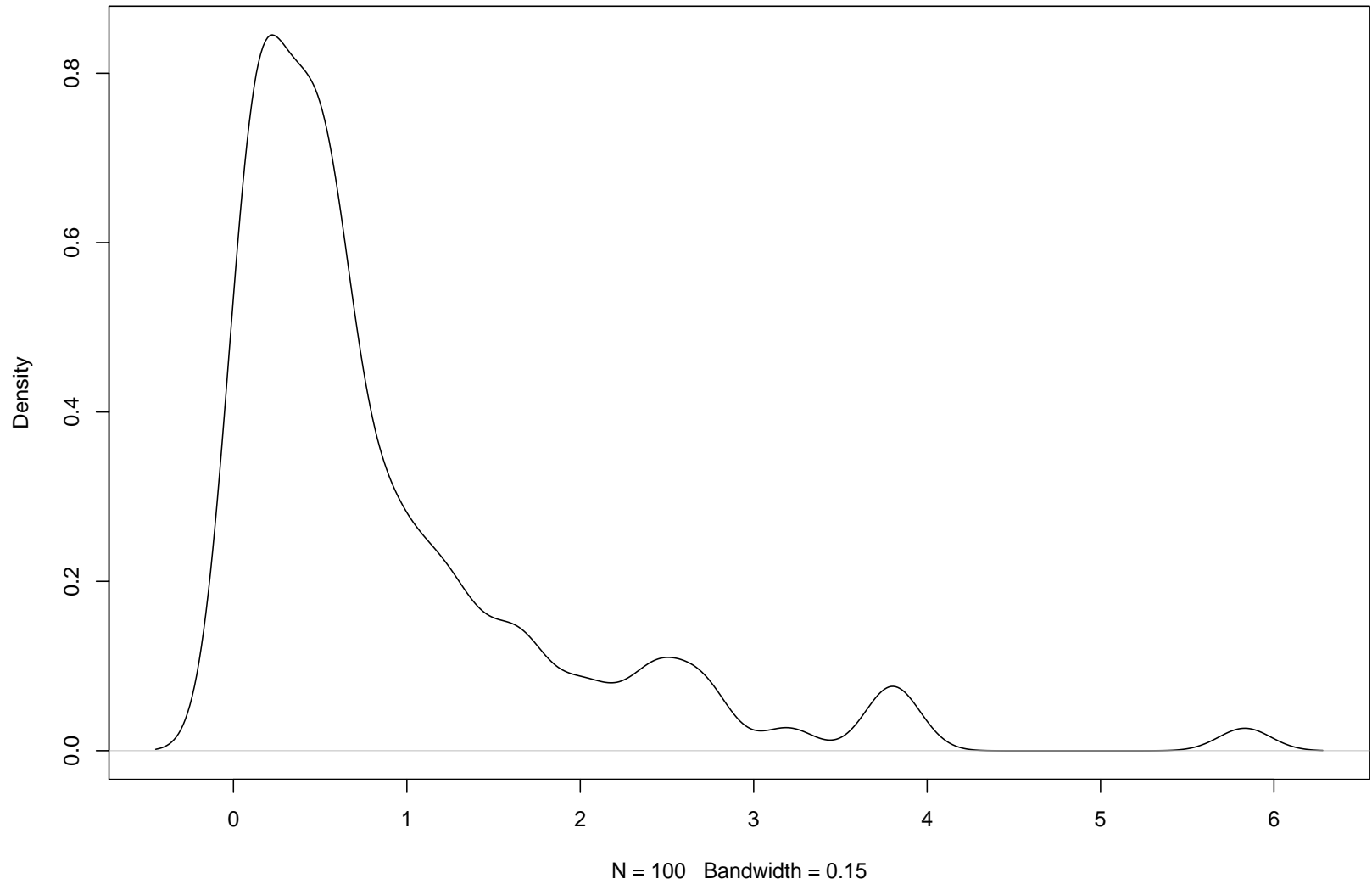


EMPIRICAL DISTRIBUTIONS

- Probability distributions must integrate to one
- Why force them into one of the above forms?
(Besides making it easier to teach the course?)
- Bumps and tails
- Kernel density estimators



EMPIRICAL DISTRIBUTIONS





LINEAR MODELS

- Recall: We wanted to understand the miss distance of our Really Deadly Missile System as a function of the threat characteristic: Threat velocity.
- Are there other threat characteristics of interest?
 1. Velocity (V)
 2. Radar cross section (C)
 3. Penetration aids (A)
 4. Pitch (P)



LINEAR MODELS

- When we only have one threat characteristic (velocity), we have

$$Y_i \sim N(\beta_0 + \beta_1 V_i, \sigma^2)$$

- Now, how to compactly represent the situation where we are interested in all 4 threat characteristics?

$$Y_i \sim N(\beta_0 + \beta_1 V_i + \cdots + \beta_4 P_i, \sigma^2)$$

- Not very compact (20 explanatory vars)



LINEAR MODELS

- Better:

$$\underline{\mathbf{Y}} \sim N(\underline{\mathbf{X}}\underline{\boldsymbol{\beta}}, \sigma^2 \mathbf{I}),$$

where \mathbf{X} contains all 4 explanatory variables and

$$\mathbf{I} = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & 0 & \ddots & 0 \\ 0 & \cdots & \cdots & 1 \end{pmatrix}$$



DISCUSSION

- We've just discussed statistical models for “data”.
- What types of things constitute “data”?
 1. Complex computer codes
 2. Physical experimental data
 3. Expert judgement
 4. Other



DISCUSSION

- While most of the discussion for the class will focus on simple parametric models, keep in mind that
 1. empirical distributions and non-parametric models are often useful, and
 2. more complicated models are used frequently, and come up naturally as different data sources are combined.