



**PSA/HFA Forum September 2010**

**Estimation Of Reliability Parameters  
For PSA Using Bayesian Analysis**

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# Overview

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PSA Context

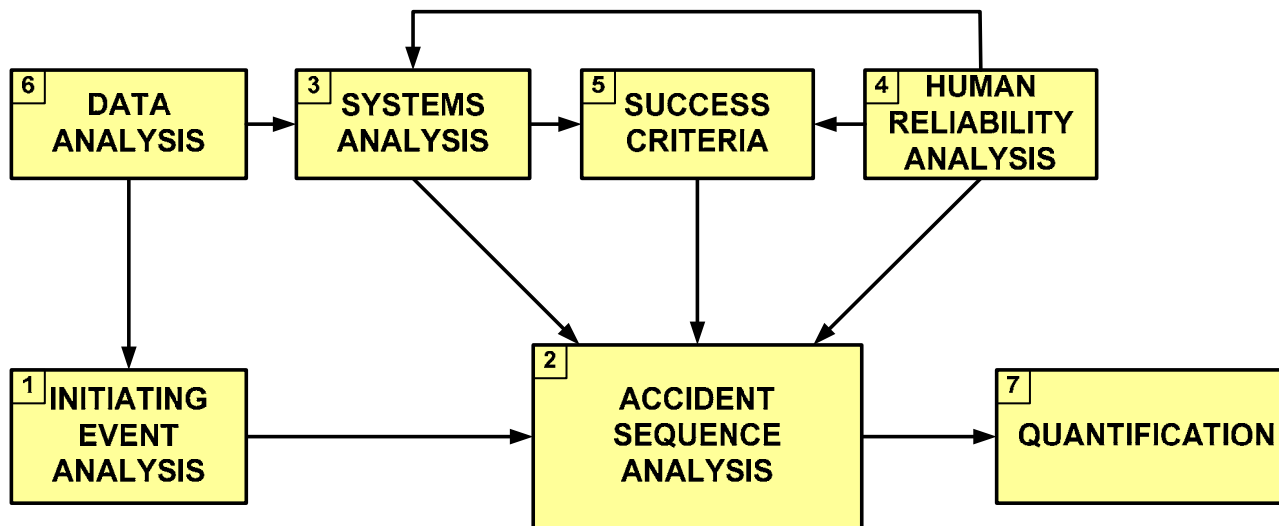
Failure Rate  
Estimation &  
Bayesian  
Analysis

Homogeneous  
& Non-  
Homogeneous  
Analysis

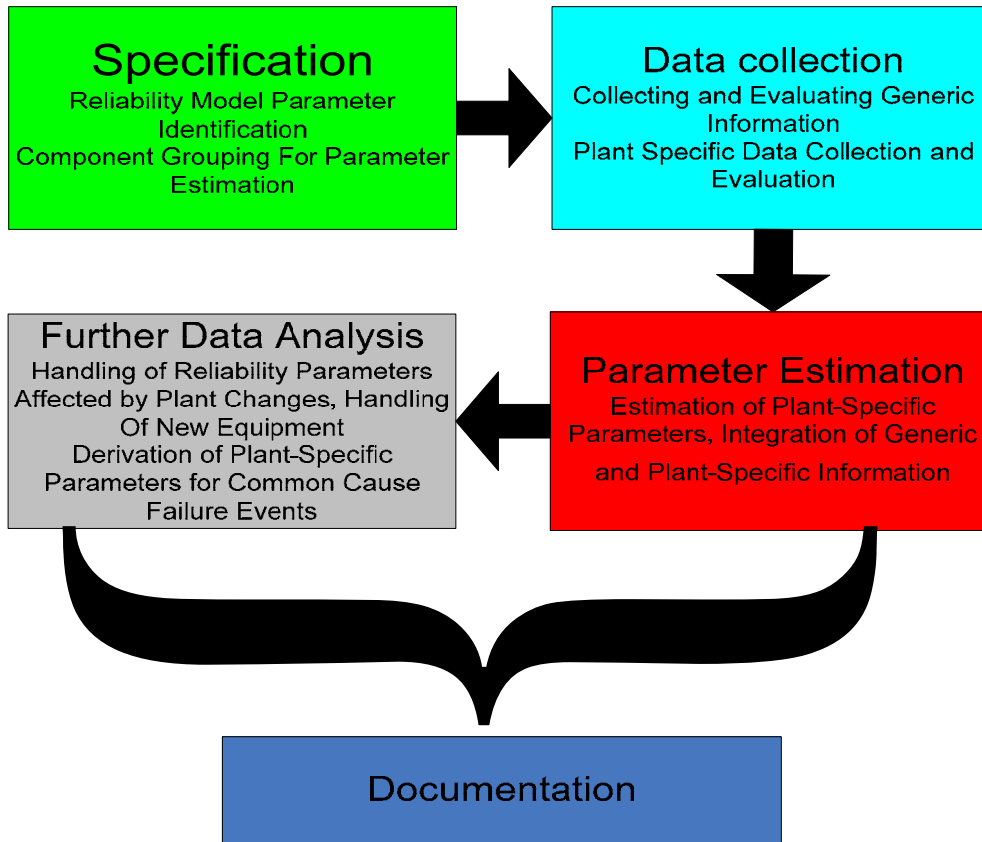
Method  
Comparison

# Probabilistic Safety Analysis In the Nuclear Industry

- ◆ Probabilistic Safety Analysis (PSA) quantifies degrees of risk at nuclear power plants
- ◆ PSA has several crucial elements



# Parameter Estimation In PSA Data Analysis



**Parameter estimation is vital in data analysis**



**Bayesian analysis is now industry standard**

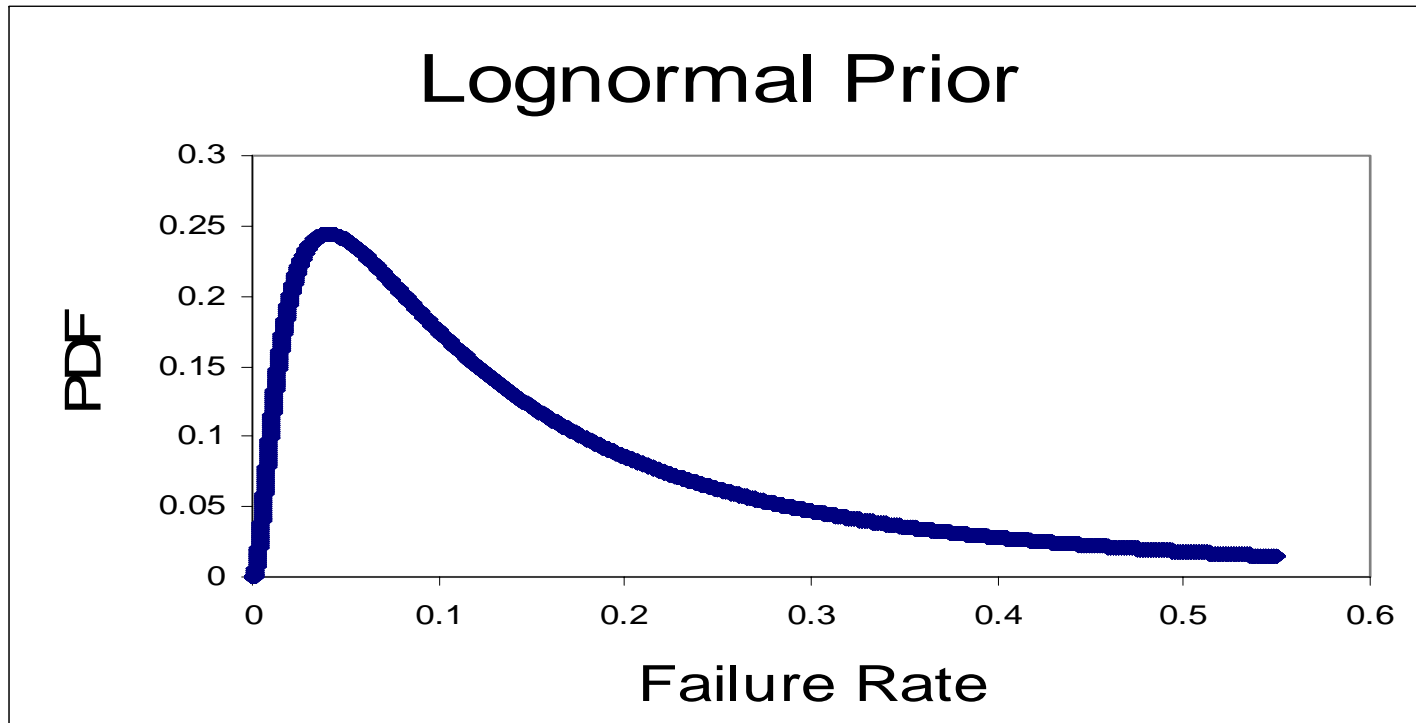
- ◆ **Components have very low failure rates (By design!)**
- ◆ **Few failures make precise estimation difficult**
- ◆ **Stochastic effects have a large influence on the estimate**
  - **Pump example:**
    - A pump operates with no failures for 999 hours
    - During the 1000<sup>th</sup> hour the pump fails
    - The MLE changes dramatically due to a single observed event

# Bayesian Analysis

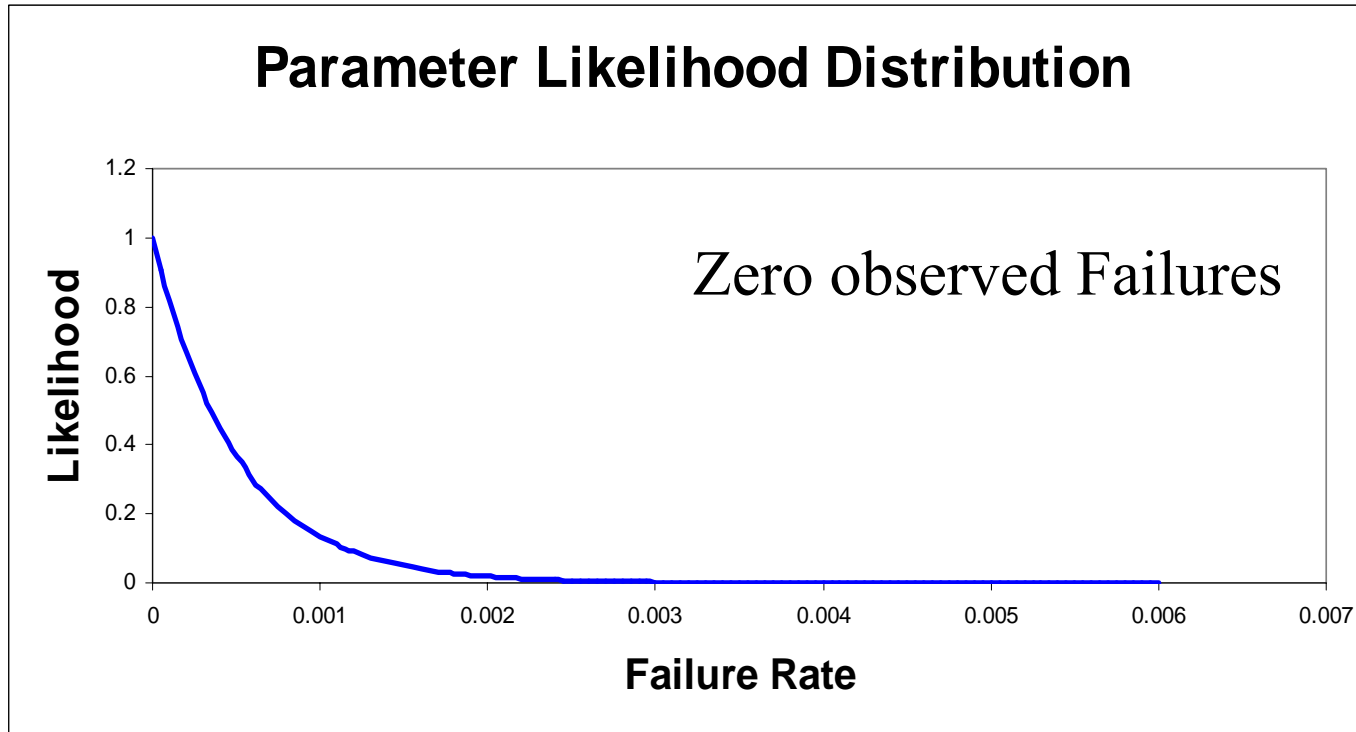
- ◆ **Parameter values are unknown variables rather than point estimates**
- ◆ **Prior estimates**
- ◆ **Likelihood and observed data**
- ◆ **Posterior Distribution**



## Prior estimate



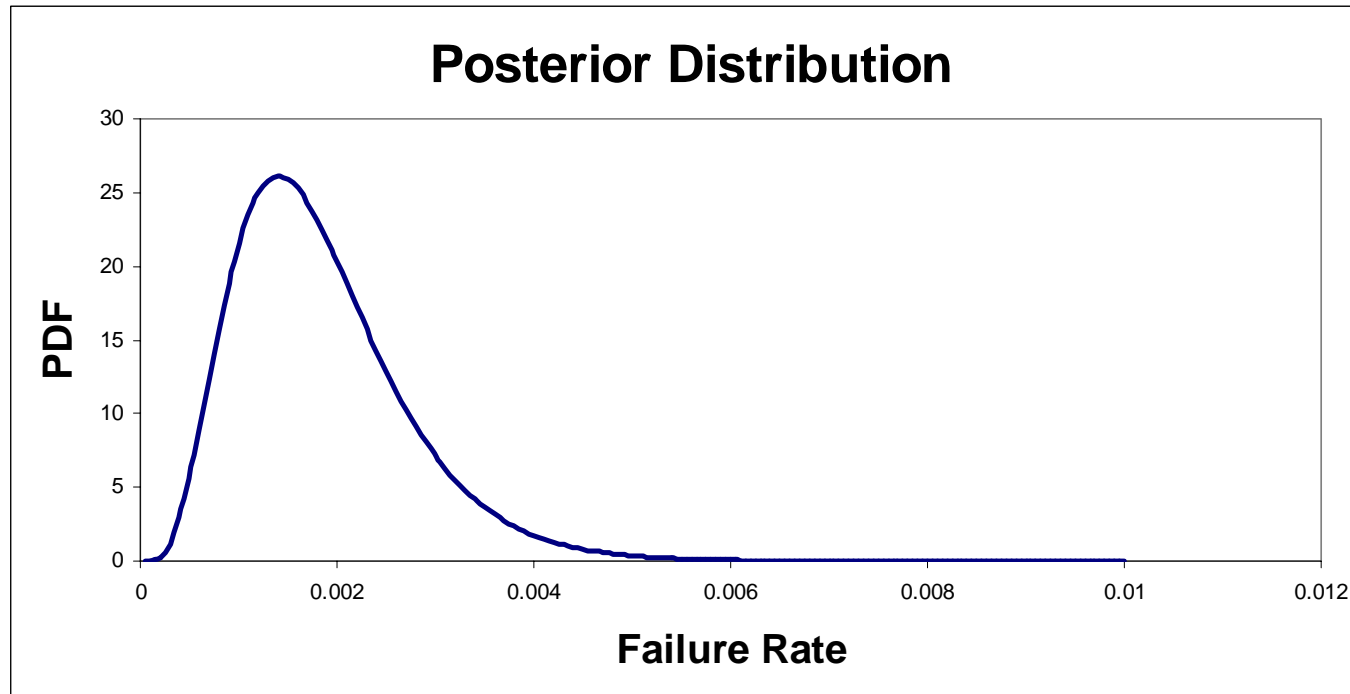
## ◆ Likelihood and observed data







## Posterior Distribution



# Bayesian Analysis

- ◆ **Parameter values are unknown variables rather than point estimates**
- ◆ **Prior estimates**
- ◆ **Likelihood and observed data**
- ◆ **Posterior Distribution**

# Bayes' Theorem

Bayes' Theorem: Relates the conditional and marginal probabilities of events A and B.

$$P(A_i|B) = \frac{P(B|A_i)P(A_i)}{P(B)}$$

where:

$P(A_i | B)$  is the posterior (or a posteriori) probability for the event  $A_i$ , i.e. the probability of  $A_i$  once B is known. The event B is the observation.

$P(B | A_i)$  is the likelihood function, i.e. the probability of the observation, B, given that the parameter takes value  $A_i$

$P(A_i)$  is the prior (or a priori) probability of the event  $A_i$  before experimentation or observation.

$P(B) = \sum P(B| A_i) P(A_i)$  is the probability of the data observations

# Homogeneous and Non-Homogeneous Bayesian Analysis

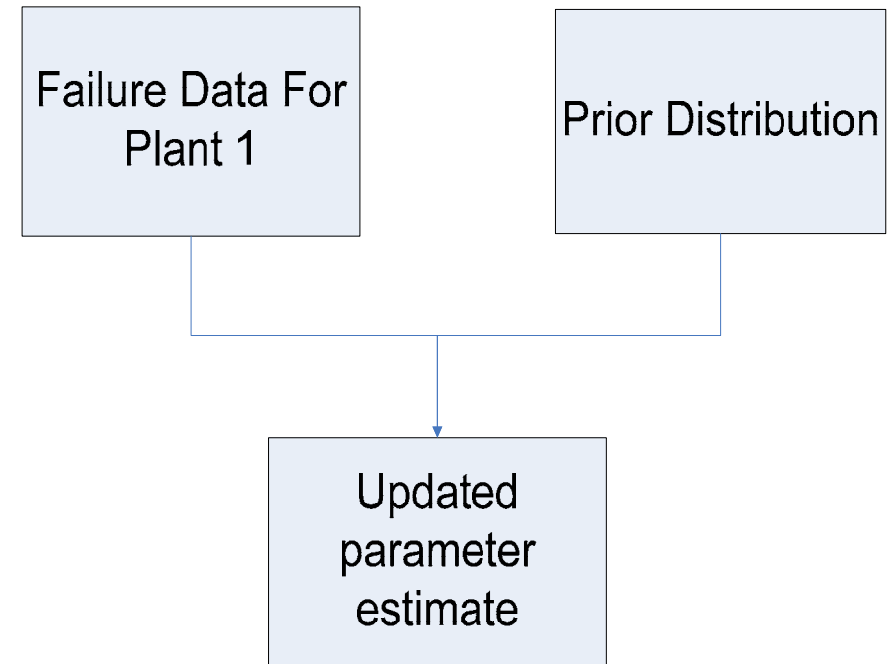
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- ◆ **Different proposed Bayesian methods**
- ◆ **Homogeneous analysis assumes equivalent sources**
- ◆ **Non-Homogeneous analysis assumes diverse sources**

# Homogeneous Analysis Structure

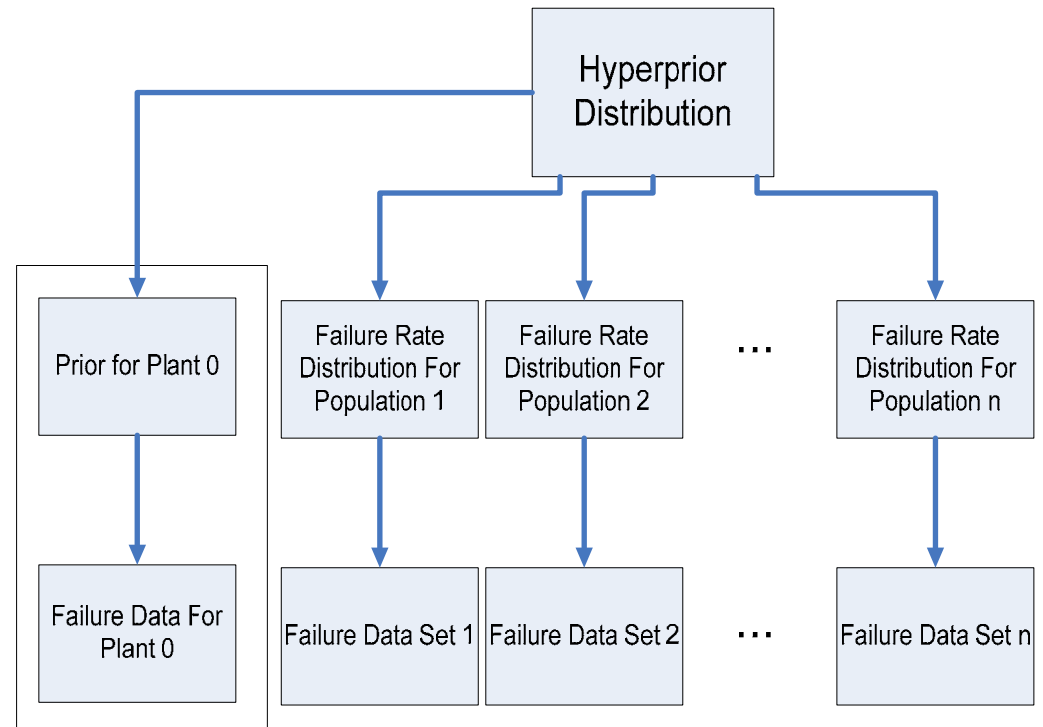
◆ **Homogeneous analysis combines observed data with prior belief**

◆ **Prior and observed data required**

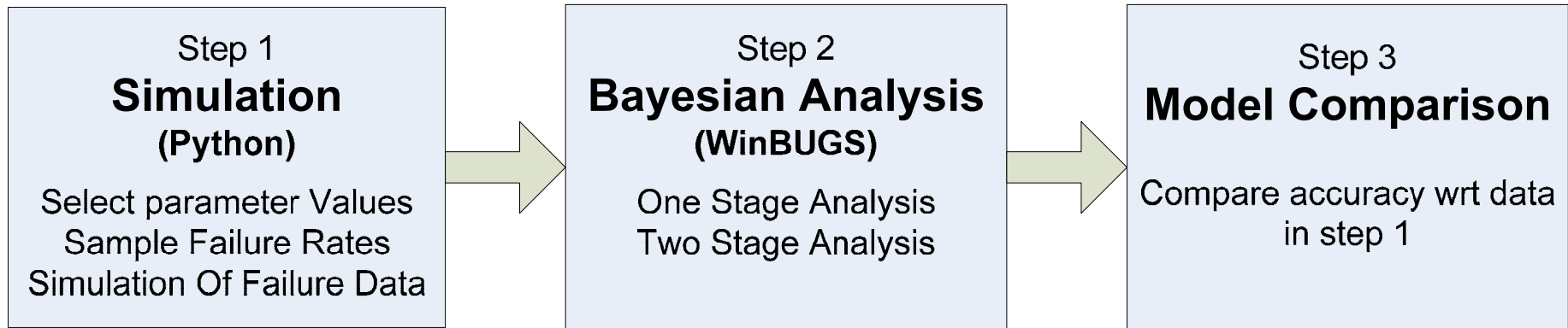


# Non-Homogeneous Analysis Structure

- ◆ **Non-homogeneous has additional conceptual layer to homogeneous analysis**
- ◆ **Additional layer of uncertainty for data observation sources**
- ◆ **Additional uncertainty introduced using a hyperprior**

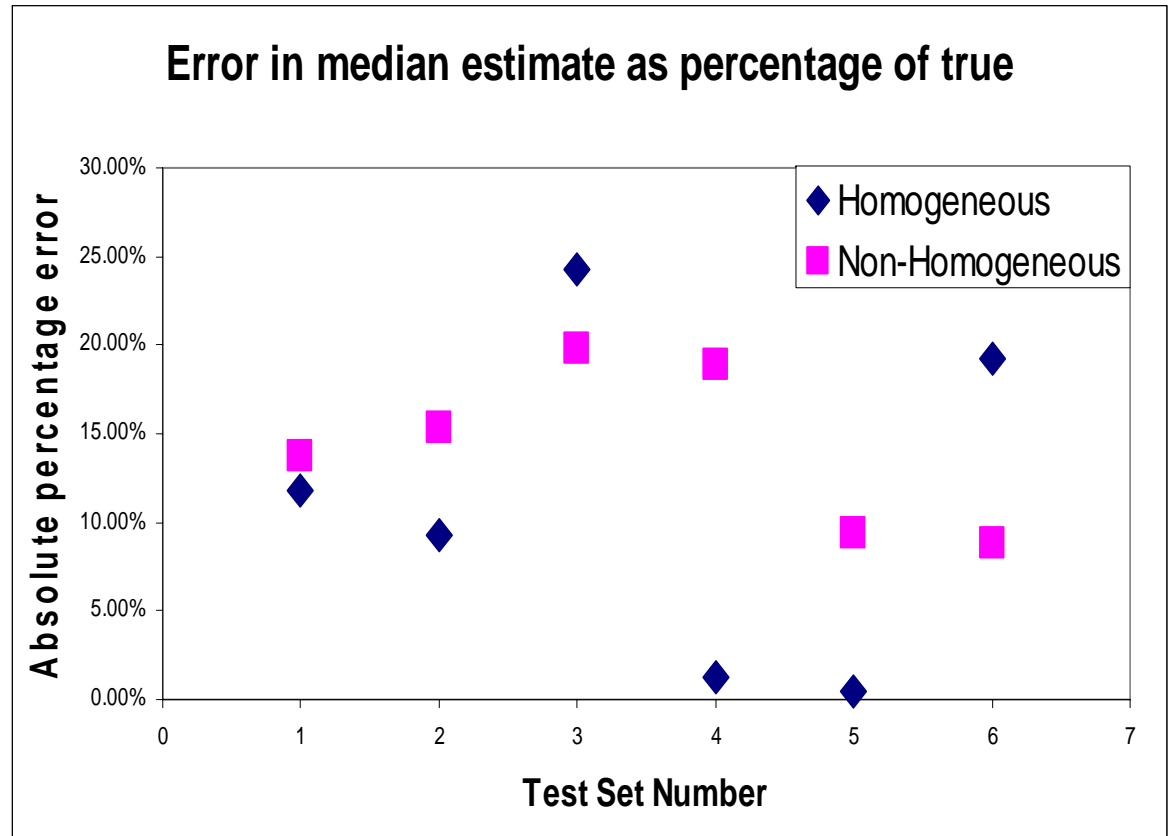


# Model Accuracy Comparison



- ◆ Simulate the failure process
- ◆ Estimate the failure rate using Bayesian models
- ◆ Accuracy comparison

- Minimal difference observed
- Stochastic variability has a large effect
- The “an extra failure occurring by chance” effect is important





## Conclusions

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- ◆ **It is difficult to discern whether homogeneous or non-homogeneous analysis is more “accurate”**
- ◆ **Homogeneous analysis may be preferred due to its simpler structure**
- ◆ **Non-homogeneous analysis provides conceptual consistency if diverse sources are used**
- ◆ **Stochastic effects cause high uncertainty in parameter estimates**

**Thanks For Listening**

**Any questions?**