

Cleaner Analysis: Quicker Decisions ***Three Examples from Government***

Decision Analysis Affinity Group
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Introduction

- **Technical Decisions for Government Projects Often Very Difficult**
 - Usually, *Lots* of Money Involved
 - Lives may be at Stake
 - Data may be Sparse
- **When a Decision Analysis is Presented, Decision Makers Often**
 - Ask Questions about Assumptions
 - Order Re-analysis with Different Assumptions
 - Generally Not Satisfied with Point Estimates
 - Afraid of Sparse Data
- **Today will Share Experiences from Three Government Project Decisions for which *Clean Analyses* Led to *Quick Decisions***

Decision Problem #1: NASA ISS O₂ Sensor Drift

- **On the International Space Station (ISS), The Extra-Vehicular-Activity (EVA) O₂ Sensor Measurements Drifting**
 - **Sensor Accuracy Requirement: $\pm 6\text{mmHg}$ for 270 Days post Calibration**
 - **Errors $> 6\text{mmHg}$: Astronaut May Suffer Bends during EVA**
 - **Errors $< -6\text{mmHg}$: Astronaut May Suffer Oxygen Toxicity**
 - **Either may result in *Death* of Astronauts**
- **NASA Faced with Either**
 - **Halting ISS EVA's Until Sensor Redesign, Testing, and Deployment**
 - **Or, Compensating for the Error Drift to Reduce the Risk**
- **Drift Compensation Results were *not Convincing***



Decision Problem #2: USCGC C130 Cooling Turbine PM

- **Cooling Turbine Provides Cooling and Pressurization to the C130 Crew**
- **Failure in Service**
 - **Loss of Cooling, but More Important, Loss of Cabin Pressurization**
 - **Smoke, Loud, Crew Must Secure**
 - **Mission Compromised**
- **Costs**
 - **Replacement: \$30,000**
 - **Refurbishment: \$500**
- **Most Cost Effective Preventative Maintenance Interval?**



Decision Problem #3: NASA ISS Bone Fracture Risks

- **On-Orbit Astronaut Bone Fractures could have Severe Consequences**
 - To the Astronaut
 - To the Mission
- **Very Low Probability Event – No Astronaut has Ever Broken a Bone during a Mission in History**
- **Risk Questions**
 - *What is the Risk of Bone Fracture for Long Mars Missions?*
 - *How Much will the Risk Increase if International Space Station Missions extend from 180 to 365 Days?*



Approach Used for These Decision Problems

- **Find a Meaningful Decision Discriminator**
 - **Some Physical Quantity or Consequence**
 - **More Importantly, One the Decision Maker *Understands and will Use* to Make the Decision**
- **Gather the Available Data**
- **Use Bayesian Methods**
 - **Use a Most General Model for Data, Express Decision Discriminator in terms of Model Parameters**
 - **Use Non-Informative or Pseudo-Ignorance Priors**
 - **Formulate Joint Posterior Uncertainty Distribution for Model Parameters**
- **Sample Joint Posterior Uncertainty Distribution using *Markov Chain Monte Carlo (MCMC)* Methods**

An Aside: Markov Chain Monte Carlo

- **Just Like Ordinary Monte Carlo, Except for Sampling Approach**
 - Ordinary Monte Carlo Uses *Built-in* Samplers for *Recognizable* Models
 - Usually Only Univariate Samplers Available
 - Possible Exception, Multivariate Normal
 - ***MCMC* uses a Markov Chain to Sample a Density Function**
 - Any Density Function – not Restricted to Recognized Built-in Models, *and Any Dimension!*
 - Any Combination of Discrete and Continuous, One-sided, Two-sided, and Interval Random Variables
- **All that is Required to Use *MCMC* is an Analytical Expression for the Density Function**

Continuing with the Approach

- **Use *MCMC* Samples for the Joint Posterior Uncertainty Distribution for Parameters to Develop Samples of the Uncertainty Distribution for the Decision Discriminator, based on the Data**
- **Parameterize if Necessary**
- **Present *Uncertainty Distributions for the Decision Discriminator* for All Alternatives**

Notes:

The Available Data

- **Most of these Tough Decision Problems Have Few *if Any* Event Data**
 - **Risks of Financial Loss Should be Small**
 - **Risks of Failure Should be Very Small**
 - **Risks of Loss of Life Should be Tiny**
- **May have Plenty of *Censored* Data**
 - **Observations that Event or Loss has Not Occurred**
 - **Classical Statistical Approaches Almost Always Ignore**
 - **Resulting Bayesian Posterior Formulations Almost Always Analytically Intractable**
- **May have Outliers – *Or, Maybe Not***
 - **May be Just one of those Rare Events**
 - **Should *never* Ignore Outliers**

Notes:

Bayesian Methods

- Decision Theory/Analysis has Long Historical Basis using Bayesian Methods
- Select the Most General Model Possible for Data
 - One-sided Data: *Weibull*
 - Two-sided Data: *Non-central t*
 - Interval Data: *Beta*
- Use of Non-Informative, or Jeffreys', or Ignorance, or Reference Priors Obviates Questionable Assumptions
 - Usually Produces Analytically Intractable Joint Posterior
 - Forced to Use *MCMC*
 - To Achieve Markov Chain Stability, Sometimes Must Wisely Truncate the Ignorance Prior – *Pseudo-Ignorance Prior*
- Bounds Results Consistent with Information Theory

Avoids Some Assumptions

Notes:

Using MCMC Joint Samples to Obtain Decision Discriminator Uncertainty Model Samples based on the Data

- **Fairly Simple Process: Evaluate Decision Discriminator at Joint MCMC Samples of Parameters**
- **What this Accomplishes**

$$pd(D | data) \propto \iiint_{\text{parameter domains}} D(\text{params}) p(\text{params} | data) dp(\text{params})$$

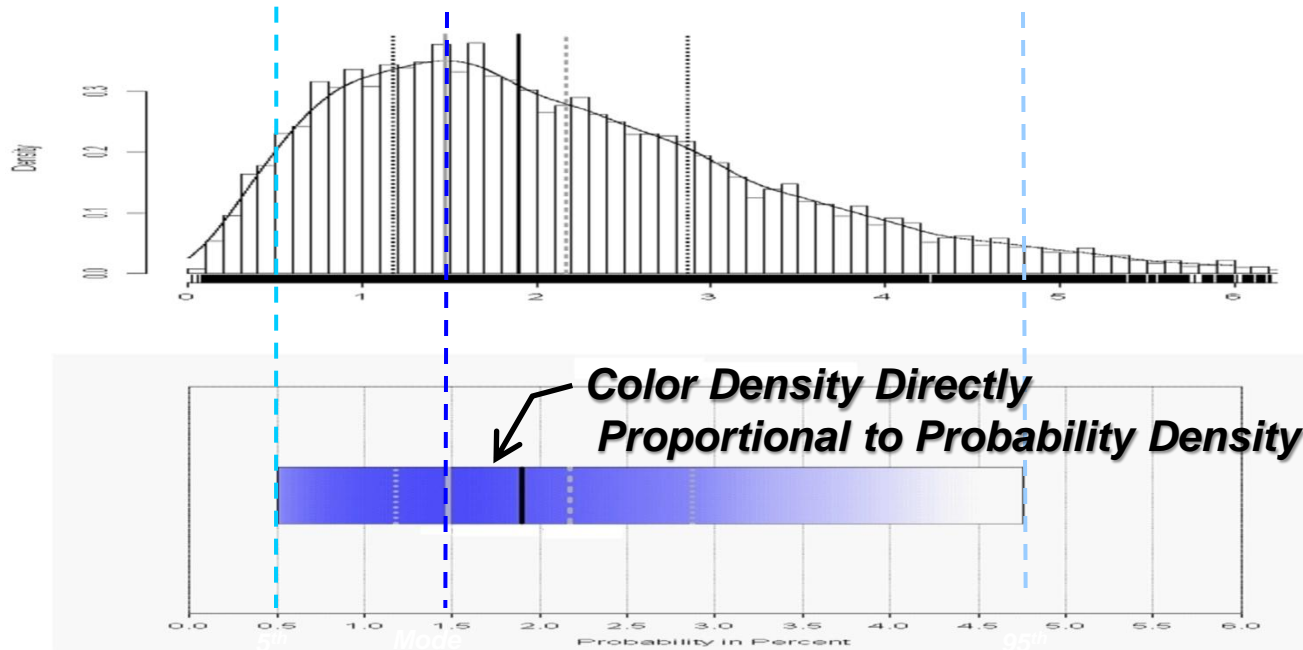
- **Performs the Required Marginalization Integrals**
- **Produces Samples of the Uncertainty Model for the Decision Discriminator**

If Needed, Parameterize

- **For Continuous Alternatives: Parameterize Decision Discriminator Uncertainty Distributions as Function of Alternative**
- **For Data with Covariates: Parameterize Decision Discriminator Uncertainty Distributions as Function of Covariates**
- **Simple, Merely Requires CPU Time**
- **Avoids a lot of Decision Maker *What if* Questions, as well as a lot of Analysis Repeats**

Alternative Distribution Presentations

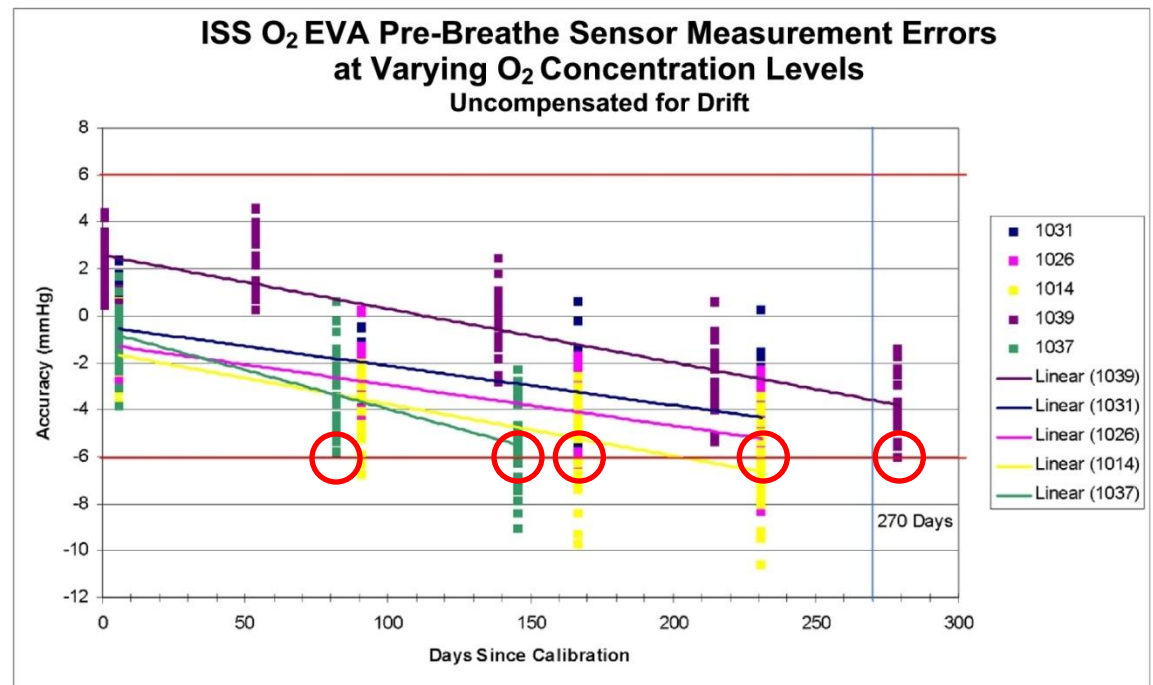
- For Discrete Alternatives, Modified Bar Charts Work Well for Risk Comparisons



Now, Decision Problem #1

Observed Sensor Errors

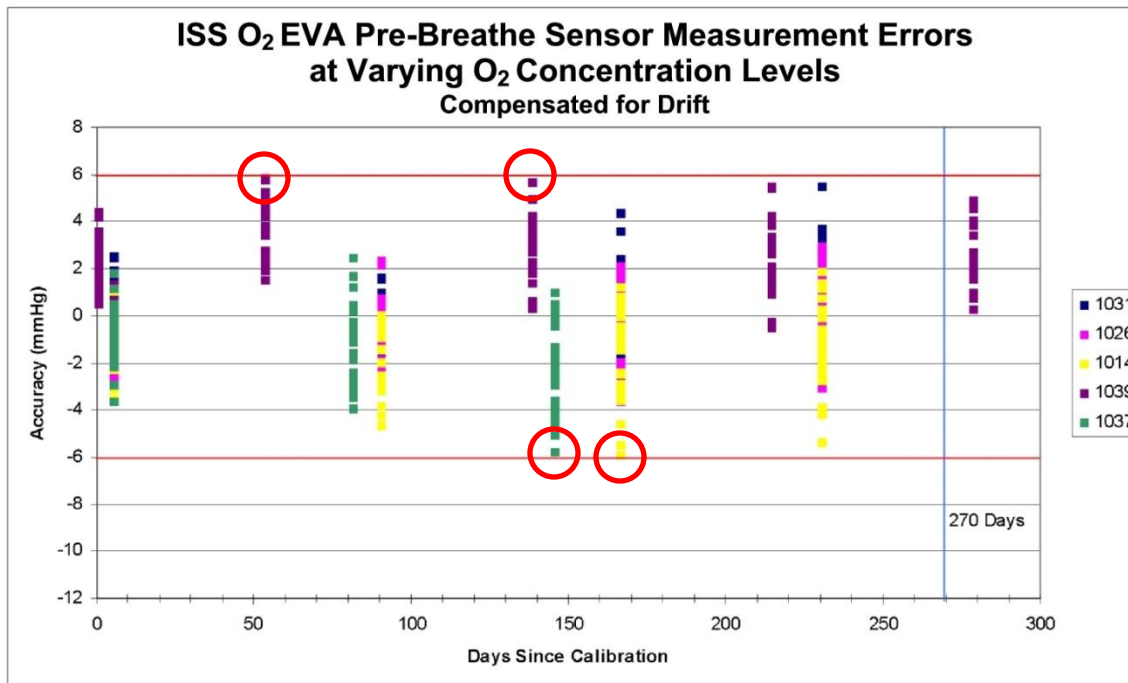
- Linear Least Squares Used to Look at Drift for Five Sensors
- All Appeared to Drift in Same Direction, with Similar Rates
- Compensation for Drift Might Reduce the Risk Enough



Compensation Scheme: Use Least Squares on All Data to Estimate Slope and Intercept, and Remove from Sensor Measurements

Sensor Errors After Drift Compensation

- Unacceptable Drift Errors Occur *Even Earlier!*
- Did the Risk *Actually Increase?*
- What was the Risk *without* Drift Compensation?
- No Answers, *No Decision!*



Decision Analysis

- **Decision Discriminator: Risk of Exceeding e_{max} ($\pm 6\text{mmHg}$) at TSC = 270 days** $R(|e_s| > e_{max} | 270, \mu_0, \mu', \sigma_s) = 2 * \Phi(-e_{max} | \mu_0 + \mu' * 270, \sigma_s)$
- **Data: Preceding Slides, Before and After Compensation**
- **Bayesian Approach**
 - **Used Normal Model with Covariate for TSC since Linear Regression was Used to Compute Drift Correction Parameters**
 - **Joint Posterior with Ignorance Priors NOT Analytically Tractable, Used MCMC Sampling**

- **Decision Discriminator Uncertainty Model Transform**

$$pd(R(|e_s| > e_{max} | 270) | data)$$

$$\infty \int \int_{-\infty -\infty}^{\infty \infty}$$

**A Very Messy Function,
Won't Need It!**

$$d\mu_0 d\mu' d\sigma_s$$

- **Use Modified Bar Charts before and after Compensation**

Another Aside:

Decision Discriminator Uncertainty

- Obtain Samples by Simply Evaluating Decision Discriminator Equation at Joint Samples of Posterior
- Suppose Want to Know Assurance *Based on the Data* that Risk of Exceeding e_{max} at TSC = 270 days is Less than 5%
 - Have M Joint Posterior Samples from MCMC
 - Evaluate Decision Discriminator Equation at Each Joint Sample at TSC = 270 days – Get M Samples of Risk of Exceeding e_{max} at TSC = 270 days Based on the Data (for Modified Bar Charts)
 - Count Number of Risk Samples < 0.05 and divide by M

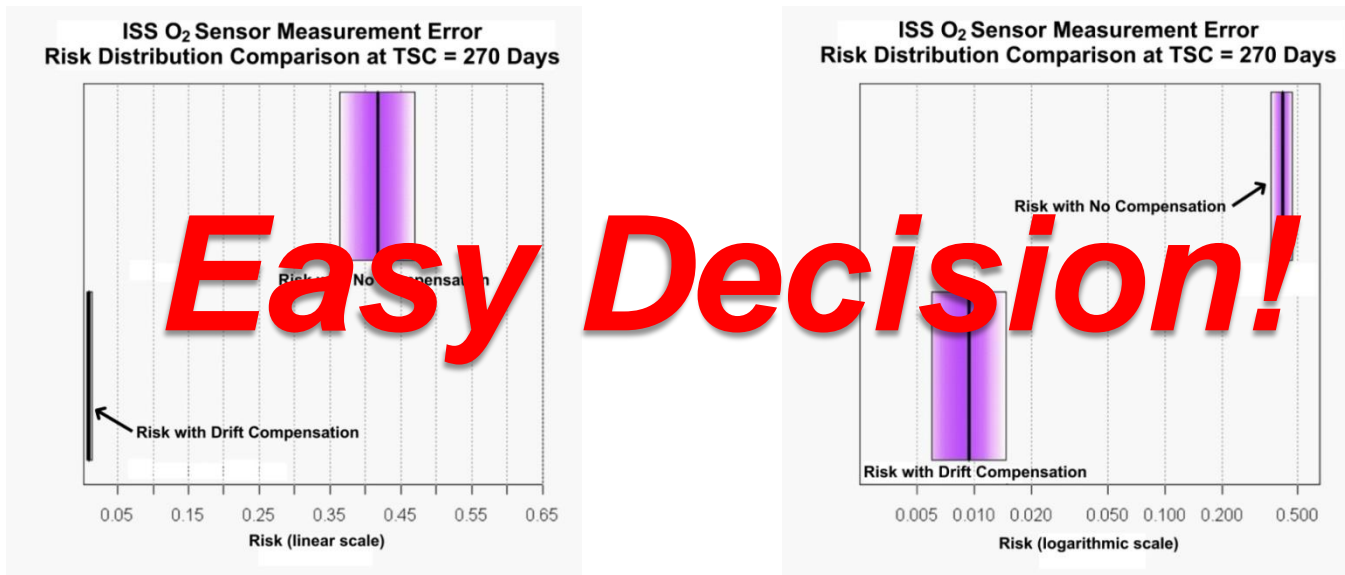
$$P\left(R(|e_s| > e_{max} \mid \text{TSC} = 270) < 0.05 \mid data\right)$$

$$= \frac{\sum_{i=1}^M \left[\begin{array}{l} 1 \mid 2 * \Phi(-e_{max} \mid \mu_{0i} + \mu'_i * 270, \sigma_{si}) < 0.05 \\ 0 \mid 2 * \Phi(-e_{max} \mid \mu_{0i} + \mu'_i * 270, \sigma_{si}) \geq 0.05 \end{array} \right]}{M}$$

M

Risk Assessment Results

- Obtained 10,000 Joint MCMC Samples of μ_0 , μ' , and σ_s for Covariate Data With and Without Drift Compensation
- Used to Obtain Risk Samples for both at TSC = 270 days



- 90% Certain Based on the Data, Risk of Exceeding e_{max} without Drift Compensation within 270 Days Between 36% and 46%
- 95% Certain Based on the Data, Risk of Exceeding e_{max} with Drift Compensation within 270 Days is less than 1.5%

Decision Problem #2

The C130 PM Problem

- **60:1 Cost Ratio, Replace:Maintain**
- **Only Had Five Failure Data: 463, 538, 1652, 1673, and 2462 flight hours**
- **Only Had One Survivor Datum: 96 flight hours**
- **What PM Interval to Select?**
- **USCG Decision Makers *Paralyzed***

Decision Analysis

- **Decision Discriminator: CS_{tpm} – Cost Savings per Flight Hour in performing Preventative Maintenance at the Interval of t_{pm} flight hours over Allowing Failures in Service**

$$CS_{tpm} = \left(\frac{C_{rep}}{\eta} \right) \gamma \left(\frac{\beta - 1}{\beta}, \left(\frac{t_{pm}}{\eta} \right)^\beta \right) - \left(\frac{C_{pm}}{t_{pm}} \right) * e^{-\left(\frac{t_{pm}}{\eta} \right)^\beta}$$

- **Data: Preceding Slides, 5 Failures Events, One Survivor**
- **Bayesian Approach**
 - Used Weibull Model
 - Posterior with Ignorance Priors *NOT* Analytically Tractable, Use *MCMC* Sampling
- **Decision Discriminator Uncertainty Model Transform**

$pd(CS_{tpm})$ **A Very Messy Function, Not Analytically Integrable!** $\left(\frac{t_{pm}}{\eta} \right)^\beta$

- **Use Parameterization as a Function of t_{pm}**

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Cost Savings Risks Using a PM Interval - Parameterized

- Full Distributions Per Flight Hour Based Solely on The Data, Per Bird – CS_{tpm}
- Obtained by Evaluating CS_{tpm} at the Joint Posterior MCMC Samples Parameterized as a function of PM Interval in flight hours
- Plotted Only 5th, Most Likely, and 95th percentile Cost Savings Risks
- At $t_{pm} = 250$ hours, 95% Certain, based on the data, that USCG can **SAVE** at least \$17 per flight hour per bird

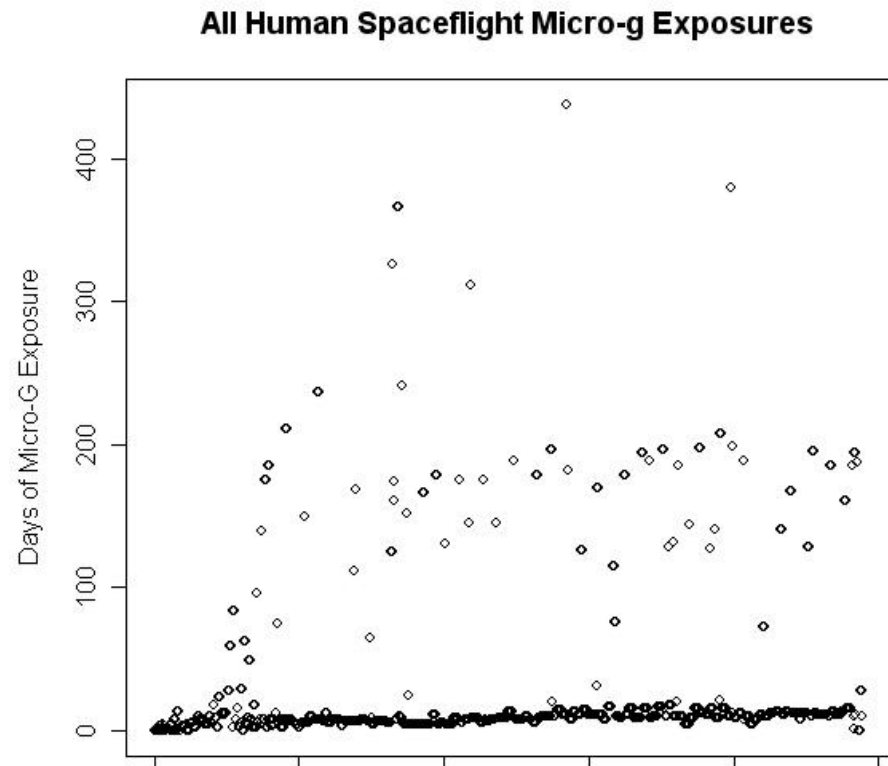
Cost Savings Risk Parameterization



Decision Problem #3

Available NASA Bone Fracture Data

- **977 Astronaut Missions of Varying Lengths (as of May 2005)**
- **No Events Observed**
 - **No Bones Broken**
 - **Did Observe 977 Mission Lengths without a Broken Bone**



Decision Analysis

- Decision Discriminator: *Risk of Bone Fracture* $R_{T_M} = 1 - e^{-\left(\frac{T_M}{\eta}\right)^\beta}$
- Data: Preceding Slides, 977 Censored Data
- Bayesian Approach
 - Used Weibull Model
 - Posterior with Ignorance Priors *NOT* Analytically Tractable, Use *MCMC* Sampling
- Decision Discriminator Uncertainty Model Transform

$pd(R_{T_M})$

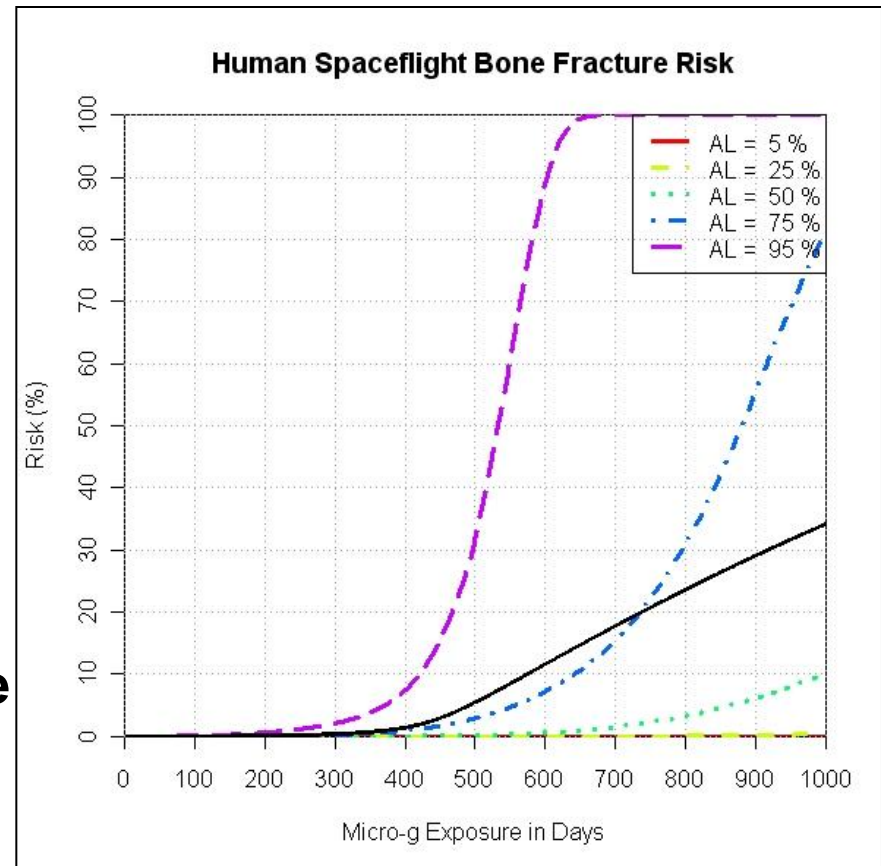
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 $\ln \beta$

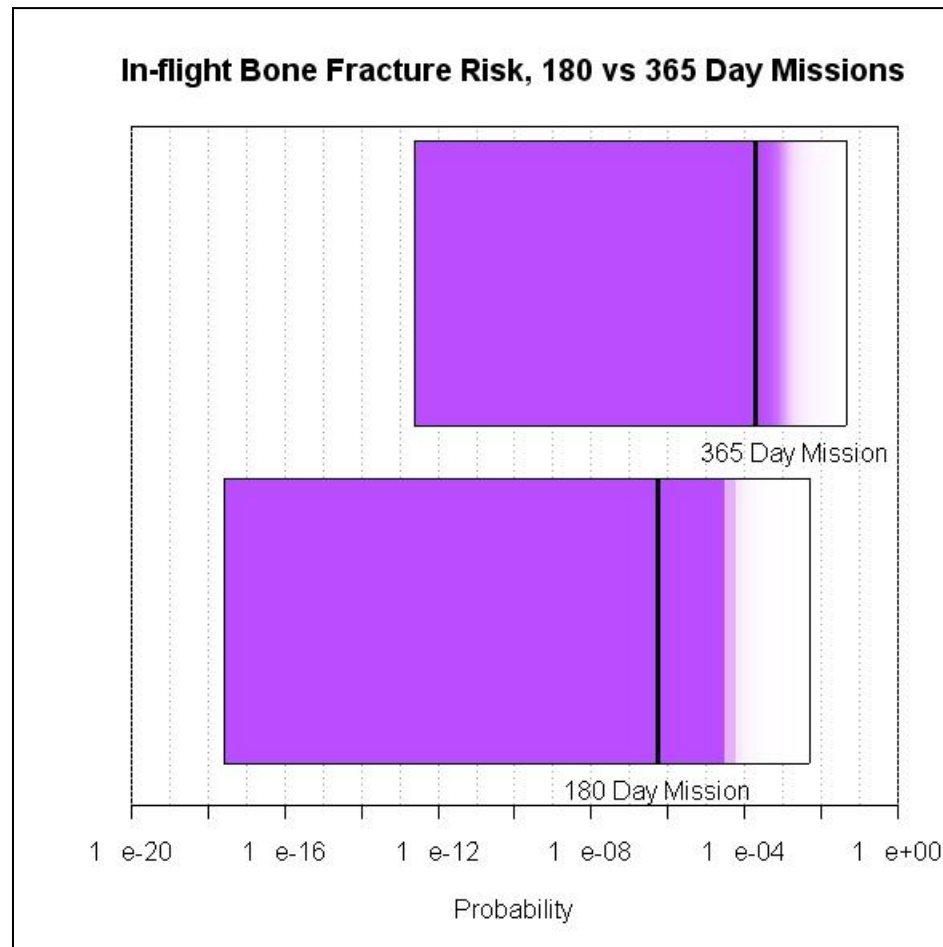
- Use Parameterization and Modified Barcharts

Risk Results Parameterized for Mission Duration

- Risk Uncertainty Distribution As a Function of Mission Duration Obtained by Evaluating Risk Equation at MCMC Samples
- Plotted Various Assurance Levels (5, 25, 50, 75, 95%)
- For Mars Missions of 270 Days – We Can be 95% Certain that Risk of fracture during the Mission is < 3%, *Based on the Information Available*



The ISS Mission Extension Question



Summary

- **The Decision Analyses Used were *Clean***
 - **Selected Meaningful and Useful Decision Discriminators**
 - **Using Ignorance or Pseudo-Ignorance Priors Limited Use of Questionable Assumptions**
 - **Used Parameterizations**
 - **Presented Uncertainty Distributions for Decision Discriminators for All Alternatives, based on the Data**
- **Decisions Made *Almost Immediately* for All Three Examples**
 - **Decision Makers were Comfortable Deciding**
 - **Saved Money in All Cases**

Conclusion

- **Have Published Papers for these 3 Examples, Contact me and I will Share**
- **Available to Help with *Tough* Decision Analysis Problems**
- **Or, I Can Teach Your Folks How to Perform Clean Decision Analysis and Achieve Quick Decision**
- ***Link with me:***
<http://www.linkedin.com/in/attwatermarkpowell>

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