ESD.33 -- Systems Engineering

Session #9
Critical Parameter Management & Error Budgeting

Dan Frey
Plan for the Session

Follow up on session #8

- Critical Parameter Management
- Probability Preliminaries
- Error Budgeting
  - Tolerance
  - Process Capability
  - Building and using error budgets

Next steps
S - Curves

Atish Banergee –

We first studied S-curves in technology strategy…The question remained why the S-curve has the peculiar shape. Well I found the answer in system dynamics. It is a general phenomenon and not restricted to technology.

It can be thought of as two curves:
1. The lower part of the curve is growth with acceleration....
2. The upper part of the s-curve is called a goal-seeking curve and can be thought of as growth with deceleration...
Trends in Compressor Performance

Evolution of Jet Engine Performance

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Critical Parameter Management

• CPM provides discipline and structure
• Produce critical parameter documentation
  – For example, a critical parameter drawing
• Traces critical parameters all the way through to manufacture and use
• Determines process capability ($C_p$ or $C_{pk}$)
• Therefore, requires probabilistic thinking
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Probability Definitions

• Sample space – a list of all possible outcomes of an experiment
  – Finest grained
  – Mutually exclusive
  – Collectively exhaustive

• Event - A collection of points in the sample space
Concept Question

• You roll 2 dice

• Give an example of a single point in the sample space?

• How might you depict the full sample space?

• What is an example of an “event”?
Probability Measure

• Axioms
  – For any event $A$, $P(A) \geq 0$
  – $P(U) = 1$
  – If $A \cap B = \emptyset$, then $P(A \cup B) = P(A) + P(B)$

For the case of rolling two dice:
$A$ = rolling a 7 and
$B$ = rolling a 1 on at least one die
Is it the case that $P(A + B) = P(A) + P(B)$?
Discrete Random Variables

- A random variable that can assume any of a set of discrete values
- Probability mass function
  \[ p_x(x_o) = \text{probability that the random variable } x \text{ will take the value } x_o \]
- Let's build a pmf for rolling two dice
  - random variable \( x \) is the total

\[
p_x(x) \quad x=10
\]
Continuous Random Variables

- Can take values anywhere within continuous ranges
- Probability density functions obey three rules

- \( P\{L < x \leq U\} = \int_{L}^{U} f_x(x) \, dx \)
- \( 0 \leq f_x(x) \) for all \( x \)
- \( \int_{-\infty}^{\infty} f_x(x) \, dx = 1 \)
Measures of Central Tendency

• Expected value
  \[ E(g(x)) = \int_a^b g(x) f_x(x) dx \]

• Mean
  \[ \mu = E(x) \]

• Arithmetic average
  \[ \frac{1}{n} \sum_{i=1}^{n} x_i \]

• Median

• Mode
Measures of Dispersion

- Variance: \( \text{VAR}(x) = \sigma^2 = E((x - E(x))^2) \)

- Standard deviation: \( \sigma = \sqrt{E((x - E(x))^2)} \)

- Sample variance: \( S^2 = \frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})^2 \)

- \( n^{th} \) central moment: \( E((x - E(x))^n) \)

- Covariance: \( E((x - E(x))(y - E(y))) \)
Sums of Random Variables

• Average of the sum is the sum of the average (regardless of distribution and independence)
  \[ E(x + y) = E(x) + E(y) \]

• Variance also sums iff independent
  \[ \sigma^2(x + y) = \sigma(x)^2 + \sigma(y)^2 \]

• This is the origin of the RSS rule
  – Beware of the independence restriction!
Concept Test

• A bracket holds a component as shown. The dimensions are independent random variables with standard deviations as noted. Approximately what is the standard deviation of the gap?

A) 0.011”
B) 0.01”
C) 0.001”

\( \sigma = 0.01" \)
\( \sigma = 0.001" \)
Uniform Distribution

• A reasonable (conservative) assumption when you know the limits of a variable but little else

\[ \sigma = \frac{(U - L)}{2\sqrt{3}} \]
Basic Application

• I have two spinners

\[ x = \text{result of blue spinner} \]
\[ y = \text{result of red spinner} \]
\[ z = x + y \]

• What are the pdfs for variables \( x, y, \) and \( z? \)

\[
P\{a < x \leq b\} = \int_{a}^{b} f_x(x) \, dx
\]

\[
\int_{-\infty}^{\infty} f_x(x) \, dx = 1
\]

\[ 0 \leq f_x(x) \text{ for all } x \]
Simulation Can Quickly Answer the Question

trials=10000; nbins=trials/1000;
x = random('Uniform',0,1,trials,1);
y = random('Uniform',0,2,trials,1);
z = x + y;

subplot(3,1,1); hist(x,nbins); xlim([0 3]);
subplot(3,1,2); hist(y,nbins); xlim([0 3]);
subplot(3,1,3); hist(z,nbins); xlim([0 3]);
Probability Distribution of Sums

• If $z$ is the sum of two random variables $x$ and $y$

\[ z = x + y \]

• Then the probability density function of $z$ can be computed by convolution

\[
p_z(z) = \int_{-\infty}^{z} x(z - \zeta)y(\zeta)d\zeta
\]
Convolution

\[ p_z(z) = \int_{-\infty}^{z} x(z - \zeta) y(\zeta) \, d\zeta \]
Convolution

\[ p_z(z) = \int_{-\infty}^{\infty} x(z - \zeta) y(\zeta) \, d\zeta \]
Central Limit Theorem

The mean of a sequence of $n$ iid random variables with

- Finite $\mu$

$$E\left(|x_i - E(x_i)|^{2+\delta}\right) < \infty \quad \delta > 0$$

approximates a normal distribution in the limit of a large $n$. 
Normal Distribution

\[ f_x(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \]

-6\sigma \quad -3\sigma \quad -1\sigma \quad \mu \quad +1\sigma \quad +3\sigma \quad +6\sigma

\begin{align*}
\rightarrow & \quad 68.3\% \\
\rightarrow & \quad 99.7\% \\
\rightarrow & \quad 1-2\text{ppb}
\end{align*}
Joint Normal Distribution

\[ p(x) = \frac{1}{(\sqrt{2\pi})^m \sqrt{|K|}} \exp\left\{ -\frac{1}{2} (x - \mu)^T K^{-1} (x - \mu) \right\} \]

- The lines of constant probability density are ellipsoids
- If the matrix \( K \) is diagonal, then the variables are uncorrelated and independent
Independence

• Random variables $x$ and $y$ are said to be independent iff

$$f_{xy}(x,y) = f_x(x)f_y(y)$$

• Or, knowledge of $x$ provides no information to update the distribution of $y$
Expectation Shift

\[ S = E(y(x)) - y(E(x)) \]

Under utility theory (DBD), \( S \) is a key difference between probabilistic and deterministic design.
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- Next steps
Error Budgets

• A tool for predicting and managing variability in an engineering system
• A model that propagates errors through a system
• Links aspects of the design and its environment to tolerance and capability
• Used for tolerance design, robust design, diagnosis…
Engineering Tolerances

• Tolerance --The total amount by which a specified dimension is permitted to vary (ANSI Y14.5M)

• Every component within spec adds to the yield (Y)
Tolerance on Position

Lead

$>25\% W$

Land
Tolerance of Form

THIS ON A DRAWING MEANS THIS

0.25 wide tolerance zone
# GD&T Symbols

## Geometric Characteristic Symbols

<table>
<thead>
<tr>
<th>Type of Tolerance</th>
<th>Characteristic</th>
<th>Symbol</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>For Individual Features</strong></td>
<td>Straightness</td>
<td>—</td>
</tr>
<tr>
<td>Form</td>
<td>Flatness</td>
<td>□</td>
</tr>
<tr>
<td></td>
<td>Circularity (Roundness)</td>
<td>○</td>
</tr>
<tr>
<td></td>
<td>Cylindricity</td>
<td>◦</td>
</tr>
<tr>
<td><strong>For Individual or Related Features</strong></td>
<td>Profile of a Line</td>
<td>⊙</td>
</tr>
<tr>
<td>Profile</td>
<td>Profile of a Surface</td>
<td>⊙</td>
</tr>
<tr>
<td><strong>Orientation</strong></td>
<td>Angularity</td>
<td>◦</td>
</tr>
<tr>
<td></td>
<td>Perpendicularity</td>
<td>⊥</td>
</tr>
<tr>
<td></td>
<td>Parallelism</td>
<td>//</td>
</tr>
<tr>
<td><strong>Location</strong></td>
<td>Position</td>
<td>○</td>
</tr>
<tr>
<td></td>
<td>Concentricity</td>
<td>◎</td>
</tr>
<tr>
<td><strong>Runout</strong></td>
<td>Circular Runout</td>
<td>†</td>
</tr>
<tr>
<td></td>
<td>Total Runout</td>
<td>†</td>
</tr>
</tbody>
</table>

*†Arrowhead(s) may be filled in.*
Multiple Tolerances

• Most products have many tolerances
• Tolerances are pass / fail
• All tolerances must be met (dominance)
Variation in Manufacture

- Many noise factors affect the system
- Some noise factors affect multiple dimensions (leads to correlation)
Process Capability Indices

- Process Capability Index
  \[ C_p \equiv \frac{(U - L)/2}{3\sigma} \]

- Bias factor
  \[ k = \frac{\mu - \frac{U + L}{2}}{(U - L)/2} \]

- Performance Index
  \[ C_{pk} \equiv C_p(1 - k) \]
• Motorola’s “6 sigma” programs suggest that we should strive for a $C_p$ of 2.0. If this is achieved but the mean is off target so that $k=0.5$, estimate the process yield.
C_p and k Determine Yield

- By definition

\[ Y_{FT} = \int_{L}^{U} p(q) dq \]

- If Gaussian

\[ Y_{FT} = \frac{1}{2} \left[ \text{erf} \left( \frac{3\sqrt{2}}{2} C_p (1 - k) \right) + \text{erf} \left( \frac{3\sqrt{2}}{2} C_p (1 + k) \right) \right] \]

This function to maps \( C_p \) and \( k \) to yield
C_p and k Determine Quality Loss

\[
\text{Quality Loss} = \frac{A_o}{[(U - L)/2]^2} \left( d - \frac{U + L}{2} \right)^2
\]

\[
\text{E(Quality Loss)} = A_o \left( k^2 + \frac{1}{9C_p^2} \right)
\]

---

Taguchi's quality loss function

ANSI's implied quality loss function
Crankshafts

• What does a crankshaft do?
• How would you define the tolerances?
• How does variation affect performance?
Printed Wiring Boards

• What does the second level connection do?
• How would you define the tolerances?
• How does variation affect performance?
$C_p$ and $k$ for the System

$C_p = 0.82$

$k = 0.08$

$Y_{FT} = 98.3\%$
Producibility Analysis

• Rolled throughput yield ($Y_{RT}$)--
  The probability that all tolerances are met

• Motorola’s approach
  \[ Y_{RT} = \prod_{i=1}^{m} Y_{FT_i} \]

• Assumes probabilistic independence

Motorola’s formula
\[ Y_{RT} = 0.983^{368} = 0.2\% \]

Hughes’ data
\[ Y_{RT} = 66.7\% \]
Surface Mount Data
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Error Sources

• Kinematic errors
  – Straightness
  – Squareness
  – Bearings
• Drive related errors
• Thermal errors
• Static loading
• Dynamics
Errors in a Linear Drive

- Lead deviation (µm)
- Cumulative lead error (µm/mm)
- Once per revolution lead error (µm)
- Nominal travel (mm)
Angular Errors

OK, so you put the error in the model. Now what will happen when the machine moves?
A Model of a Robot

[Diagram showing a robot model with labeled dimensions: 1000 mm, 500 mm, 400 mm, 300 mm, 60 mm, and a point labeled as Point p.]
## Errors in the Robot

<table>
<thead>
<tr>
<th>Error</th>
<th>Description</th>
<th>$\mu$</th>
<th>$\sigma$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\epsilon_{z1}$</td>
<td>Drive error of joint #1</td>
<td>0 rad</td>
<td>0.0001 rad</td>
</tr>
<tr>
<td>$\epsilon_{z2}$</td>
<td>Drive error of joint #2</td>
<td>0 rad</td>
<td>0.0001 rad</td>
</tr>
<tr>
<td>$\delta_{z3}$</td>
<td>Drive error of joint #3</td>
<td>$Z \cdot 0.0001$</td>
<td>0.01mm</td>
</tr>
<tr>
<td>$\epsilon_{x3}$</td>
<td>Pitch of joint #3</td>
<td>0 rad</td>
<td>0.00005 rad</td>
</tr>
<tr>
<td>$\epsilon_{y3}$</td>
<td>Yaw of joint #3</td>
<td>0 rad</td>
<td>0.00005 rad</td>
</tr>
<tr>
<td>xp$_2$</td>
<td>Parallelism of joint 2 in the x direction</td>
<td>0.0002 rad</td>
<td>0.0001 rad</td>
</tr>
</tbody>
</table>
A Model of a Robot

• The matrices describe the intended motions and the errors

\[
0^T = \begin{bmatrix}
1 & 0 & 0 & 1000 \text{mm} \\
0 & 1 & 0 & 0 \text{mm} \\
0 & 0 & 1 & 0 \text{mm} \\
0 & 0 & 0 & 1
\end{bmatrix}
\begin{bmatrix}
\cos(\Theta_1 + \epsilon_1) & -\sin(\Theta_1 + \epsilon_1) & 0 & 0 \text{mm} \\
\sin(\Theta_1) & \cos(\Theta_1 + \epsilon_1) & 0 & 0 \text{mm} \\
0 & 0 & 1 & 0 \text{mm} \\
0 & 0 & 0 & 1
\end{bmatrix}
\begin{bmatrix}
1 & 0 & 0 & 500 \text{mm} \\
0 & 1 & -x_p & 0 \text{mm} \\
0 & x_p & 1 & 60 \text{mm}
\end{bmatrix}
\]

\[
\begin{bmatrix}
\cos(\Theta_2 + \epsilon_2) & -\sin(\Theta_2) & 0 & 0 \text{mm} \\
\sin(\Theta_2) & \cos(\Theta_2 + \epsilon_2) & 0 & 0 \text{mm} \\
0 & 0 & 1 & 0 \text{mm} \\
0 & 0 & 0 & 1
\end{bmatrix}
\begin{bmatrix}
1 & 0 & 0 & 400 \text{mm} \\
0 & 1 & -\epsilon_{y_3} & 0 \text{mm} \\
0 & 0 & 1 & \epsilon_{z_3} \\
0 & 0 & 0 & 1
\end{bmatrix}
\begin{bmatrix}
1 & 0 & \epsilon_{y_3} & 0 \text{mm} \\
0 & 1 & -\epsilon_{z_3} & 0 \text{mm} \\
-\epsilon_{y_3} & \epsilon_{z_3} & 1 & -Z - \delta_{z_3}
\end{bmatrix}
\]

• Can be applied to any point on the end effector

\[
\begin{bmatrix}
p'_x \\
p'_y \\
p'_z
\end{bmatrix} = 0^T_3
\begin{bmatrix}
0 \\
0 \\
-300
\end{bmatrix}
\]
Homework #5

• Short answers on TRIZ and probability
• Error budgeting
  – Two tasks are to be done with the robot
  – Analyze the tasks
  – Discuss changes to the system
• A Matlab file is available in the HW folder just so you don’t have to re-type the matrices
Next Steps

• You can download HW #5 Error Budgetting
  – Due 8:30AM Tues 13 July

• See you at Thursday’s session
  – On the topic “Design of Experiments”
  – 8:30AM Thursday, 8 July

• Reading assignment for Thursday
  – All of Thomke
  – Skim Box
  – Skim Frey