

Goal Programming and Isoperformance

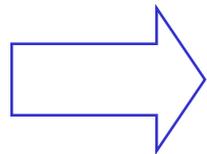
Lecture 13

Olivier de Weck

de Weck, O.L. and Jones M. B., “Isoperformance: Analysis and Design of Complex Systems with Desired Outcomes”, *Systems Engineering*, 9 (1), 45-61, January 2006

“The experience of the 1960’s has shown that for military aircraft the cost of the final increment of performance usually is excessive in terms of other characteristics and that the overall system must be optimized, not just performance”

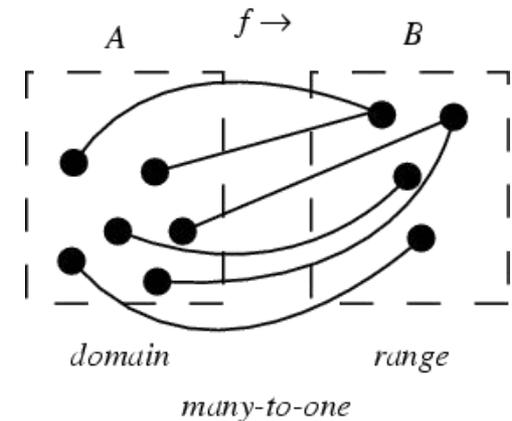
Ref: Current State of the Art of Multidisciplinary Design Optimization (MDO TC) - AIAA White Paper, Jan 15, 1991



TRW Experience

Industry designs not for optimal performance, but according to targets specified by a requirements document or contract - thus, optimize design for a set of GOALS.

- Motivation - why goal programming ?
- Example: Goal Seeking in Excel
- Case 1: Target vector \mathbf{T} in Range
= Isoperformance
- Case 2: Target vector \mathbf{T} out of Range
= Goal Programming
- Application to Spacecraft Design
- Stochastic Example: Baseball

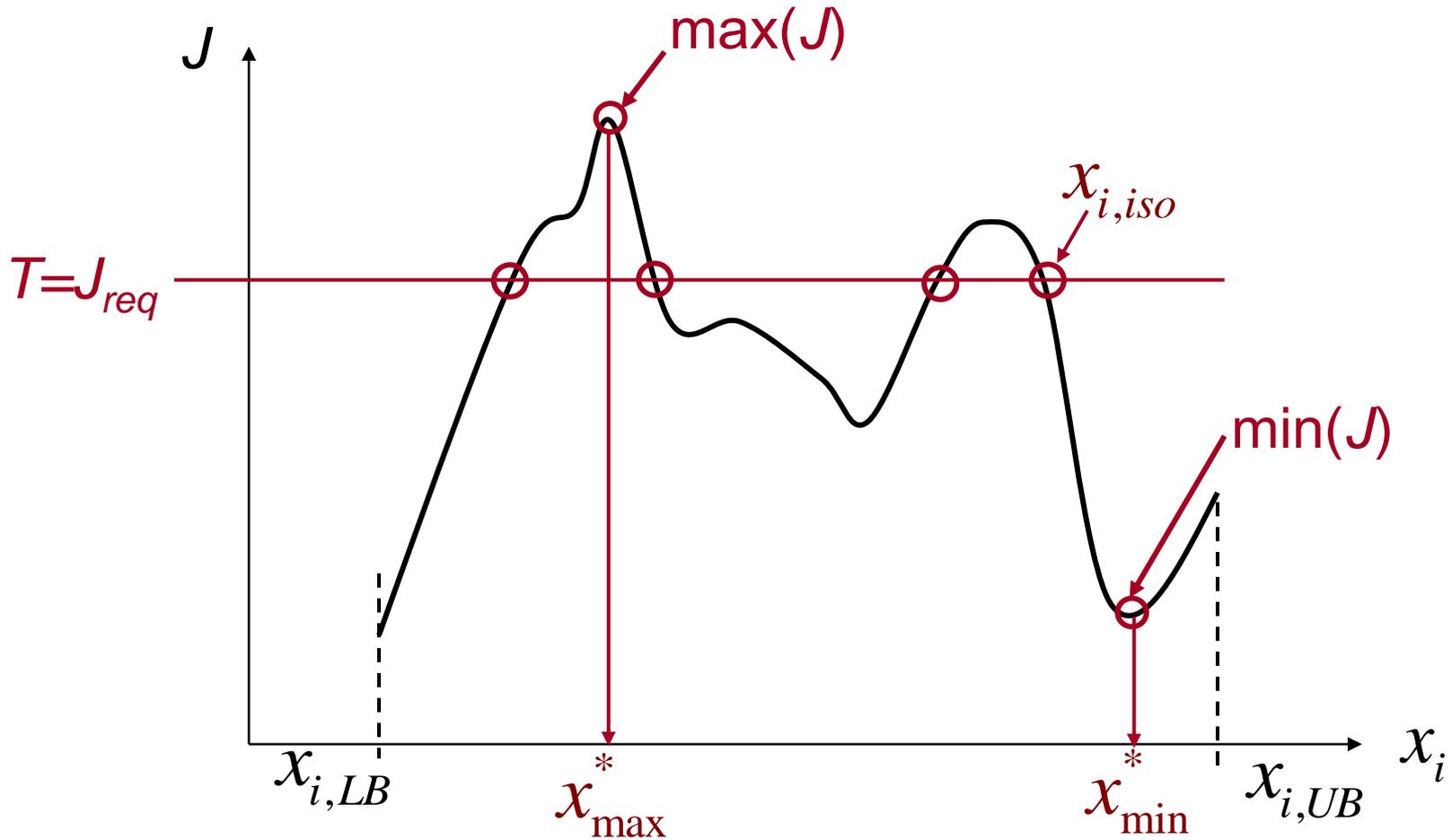
Target
Vector \mathbf{T}  \mathbf{J} 

Forward Perspective

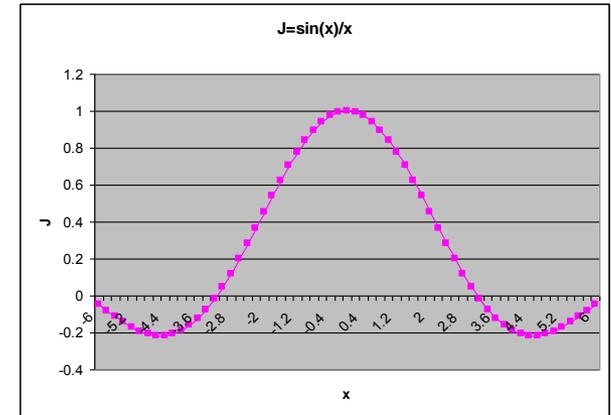
Choose \mathbf{x} \longrightarrow What is \mathbf{J} ?

Reverse Perspective

Choose \mathbf{J} \longrightarrow What \mathbf{x} satisfy this?



Excel - example



"About Goal Seek" description from Microsoft Excel.
Removed due to copyright restrictions.

 $\sin(x)/x$ - example

- single variable x
- no solution if T is out of range

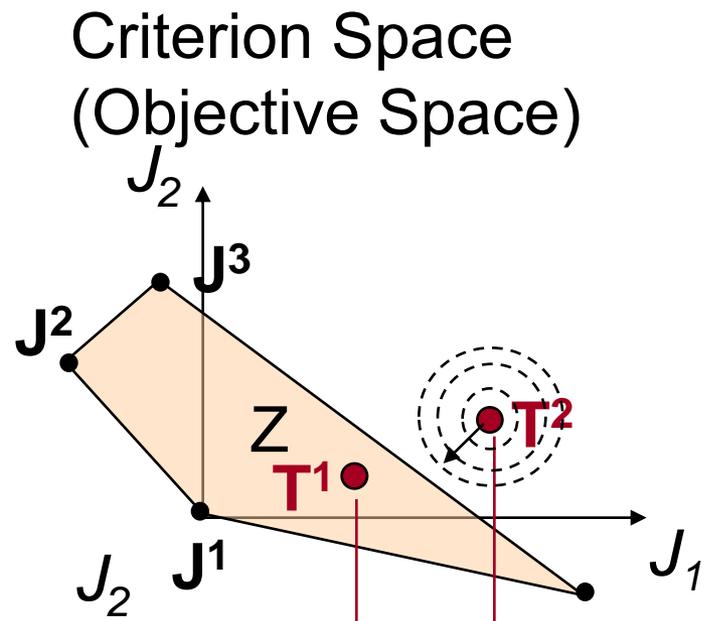
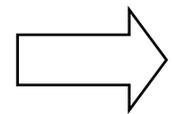
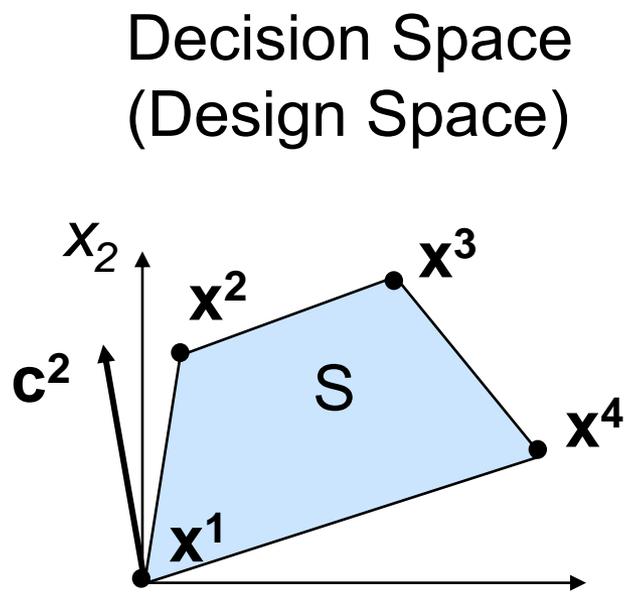
- Goal Seeking – is essentially the same as finding the set of points \mathbf{x} that will satisfy the following “soft” equality constraint on the objective:

$$\text{Find all } \mathbf{x} \text{ such that } \left| \frac{J(\mathbf{x}) - J_{req}}{J_{req}} \right| \leq \varepsilon$$

Example Target Vector:

$$J_{req}(x) = \begin{bmatrix} m_{sat} \\ R_{data} \\ C_{sc} \end{bmatrix} \equiv \begin{bmatrix} 1000kg \\ 1.5Mbps \\ 15M\$ \end{bmatrix}$$

← Target mass
← Target data rate
← Target Cost



Case 1: The target (goal) vector is in Z - usually get non-unique solutions = Isoperformance

Case 2: The target (goal) vector is not in Z - don't get a solution - find closest = Goal Programming

Non-Uniqueness of Design if $n > z$

Performance: Buckling Load

Constants: $l=15$ [m], $c=2.05$

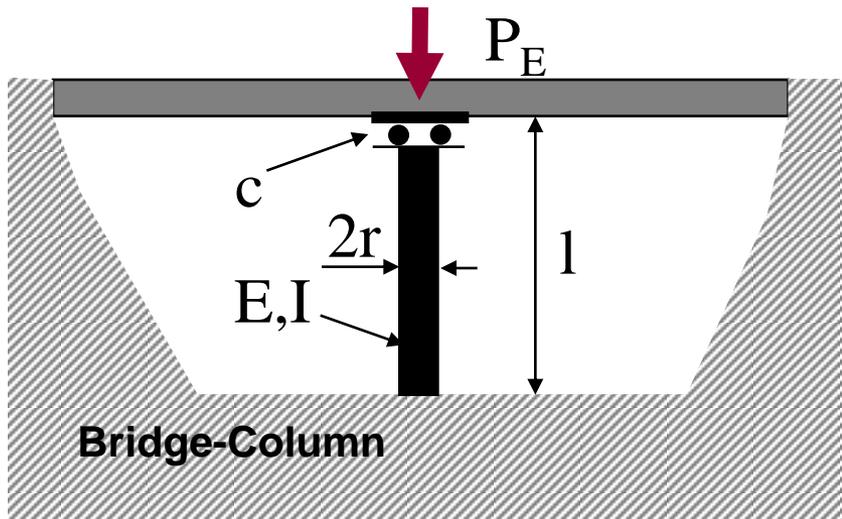
$$P_E = \frac{c\pi^2 EI}{l^2}$$

Variable Parameters: $E, I(r)$

Requirement: $P_{E,REQ} = 1000$ metric tons

Solution 1: V2A steel, $r=10$ cm, $E=19.1e+10$

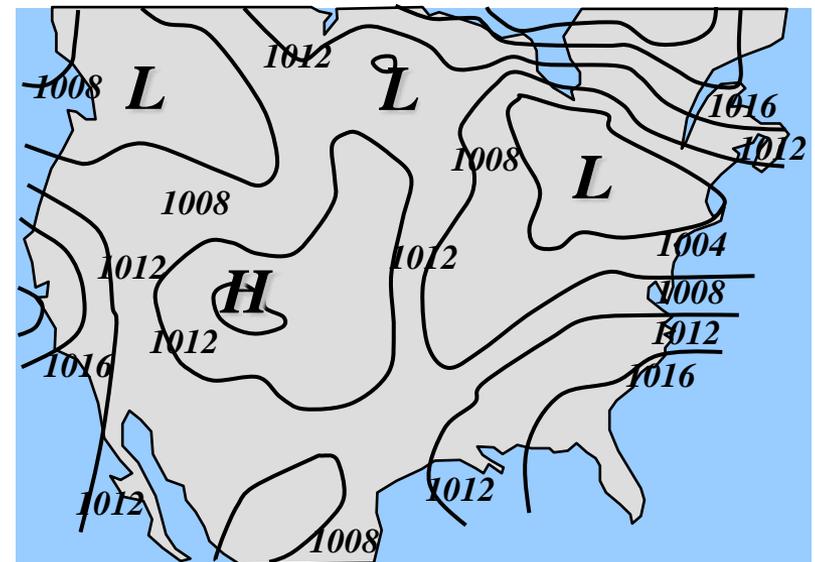
Solution 2: Al(99.9%), $r=12.8$ cm, $E=7.1e+10$



Analogy: Sea Level Pressure [mbar]

Chart: 1600 Z, Tue 9 May 2000

Isobars = Contours of Equal Pressure
Parameters = Longitude and Latitude

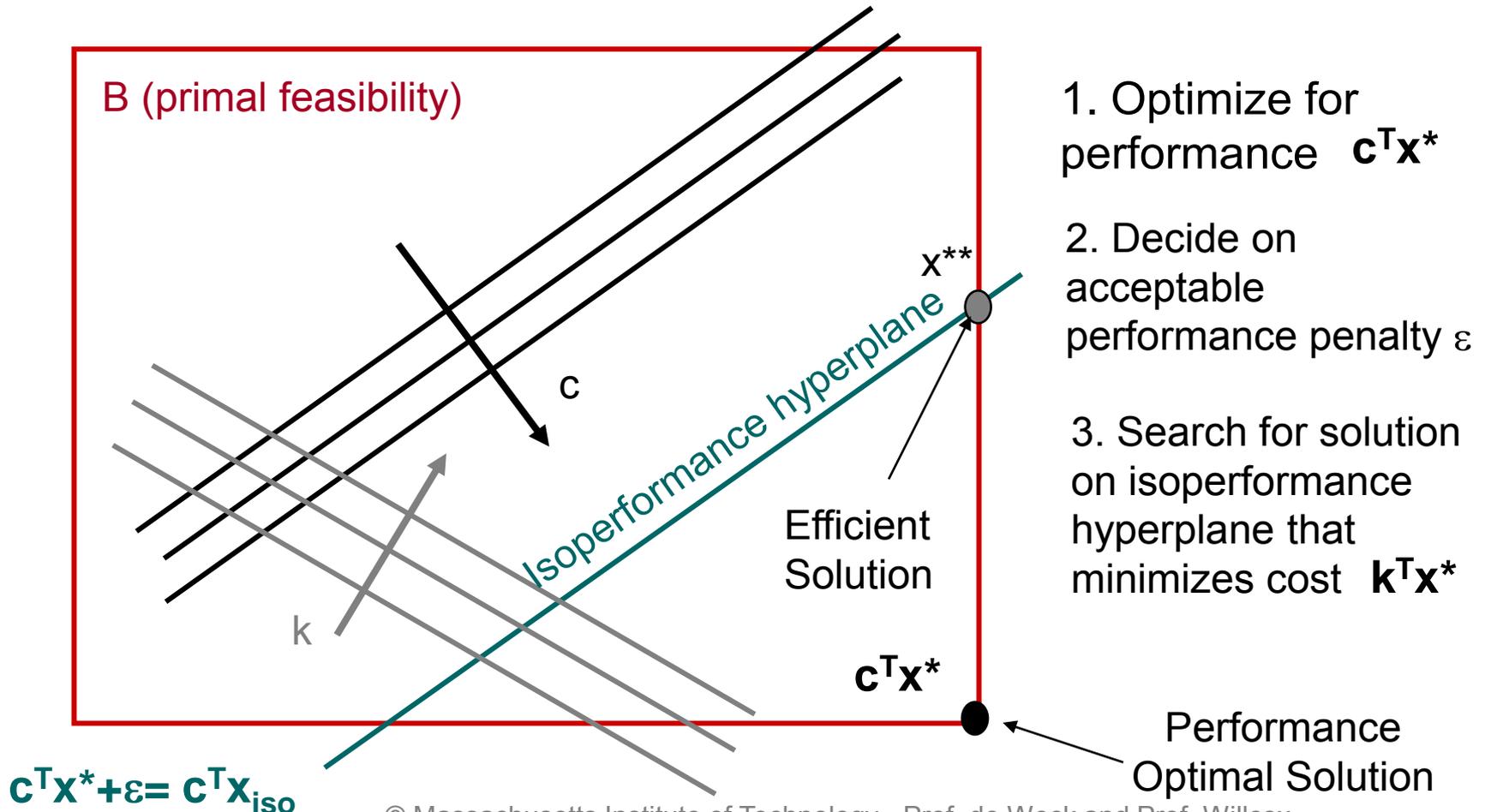


Isoperformance Contours = Locus of constant system performance
Parameters = e.g. Wheel Imbalance U_s , Support Beam I_{xx} , Control Bandwidth ω_c

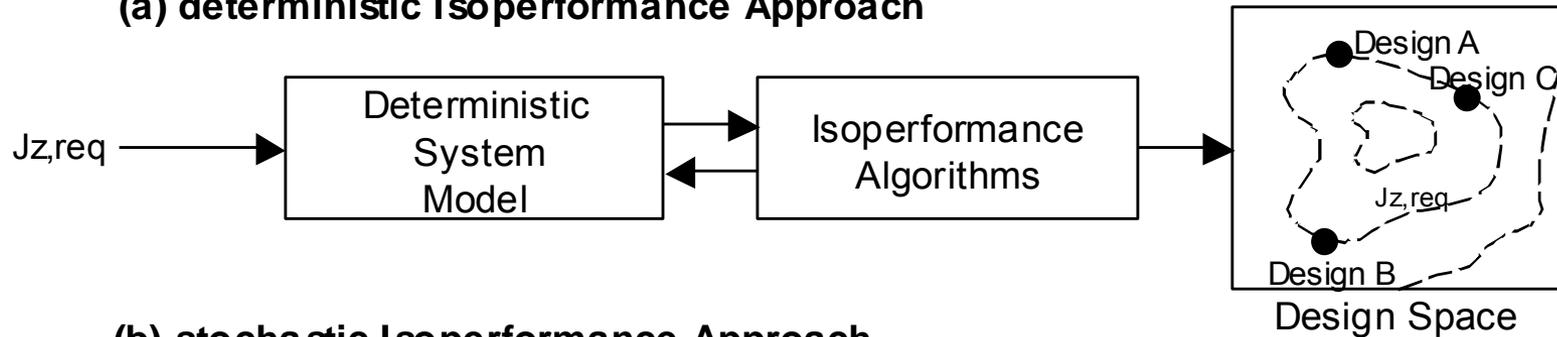
- In LP the isoperformance surfaces are hyperplanes
- Let $c^T x$ be performance objective and $k^T x$ a cost objective

$$\min c^T x$$

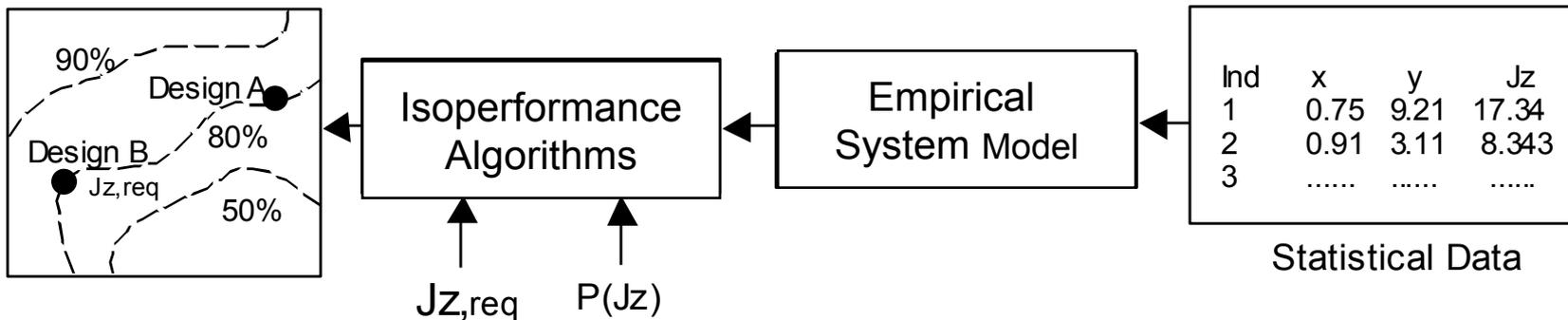
$$s.t. \quad x_{LB} \leq x \leq x_{UB}$$



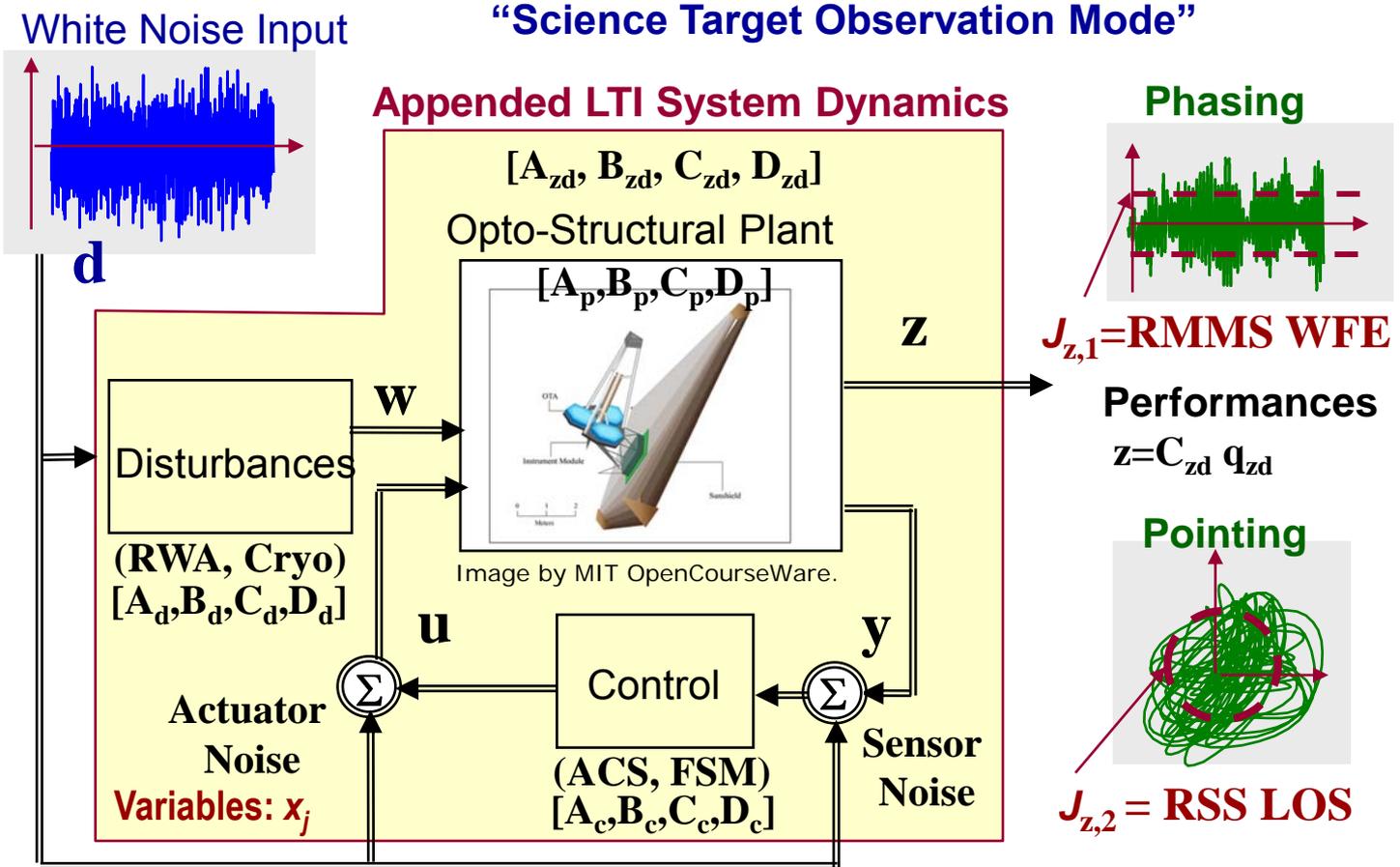
(a) deterministic Isoperformance Approach



(b) stochastic Isoperformance Approach



Courtesy of Wiley. Used with permission.



Given

$$\dot{q} = A_{zd} x_j q + B_{zd} x_j d + B_{zr} x_j r \quad \text{LTI System Dynamics}$$

$$z = C_{zd} x_j q + D_{zd} x_j d + D_{zr} x_j r, \text{ where } j = 1, 2, \dots, n_p$$

And Performance Objectives

$$J_z = F z, \text{ e.g. } J_{z,i} = \|z\|_2 = \mathbf{E} \left[z^T z \right]^{1/2} = \left(\frac{1}{T} \int_0^T z(t)^2 dt \right)^{1/2} \quad \text{RMS}$$

Find Solutions x_{iso} **such that**

$$J_{z,i} x_{iso} \equiv J_{z,req,i} \quad \forall i = 1, 2, \dots, n_z$$

Assuming $n - z \geq 1$ **and** $x_{j,LB} \leq x_j \leq x_{j,UB} \quad \forall j = 1, 2, \dots, n$ **Subject to a numerical tolerance** $\tau : \left| \frac{J_z x - J_{z,req}}{J_{z,req}} \right| \leq \frac{\tau}{100} = \varepsilon$

“Simple” Start: Bivariate Isoperformance Problem

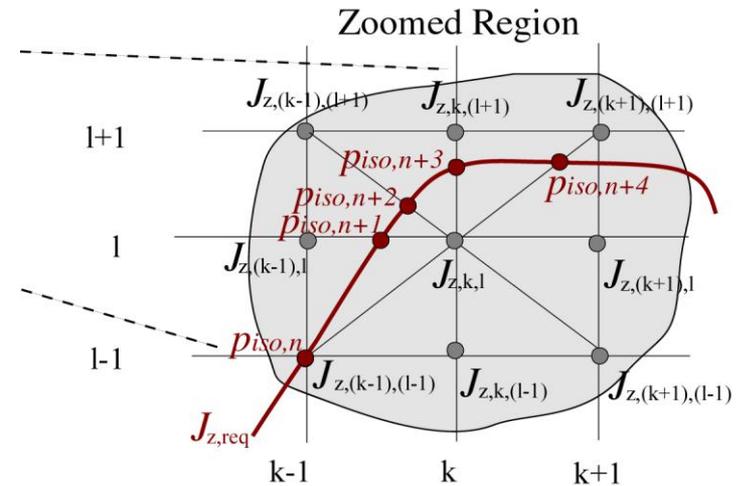
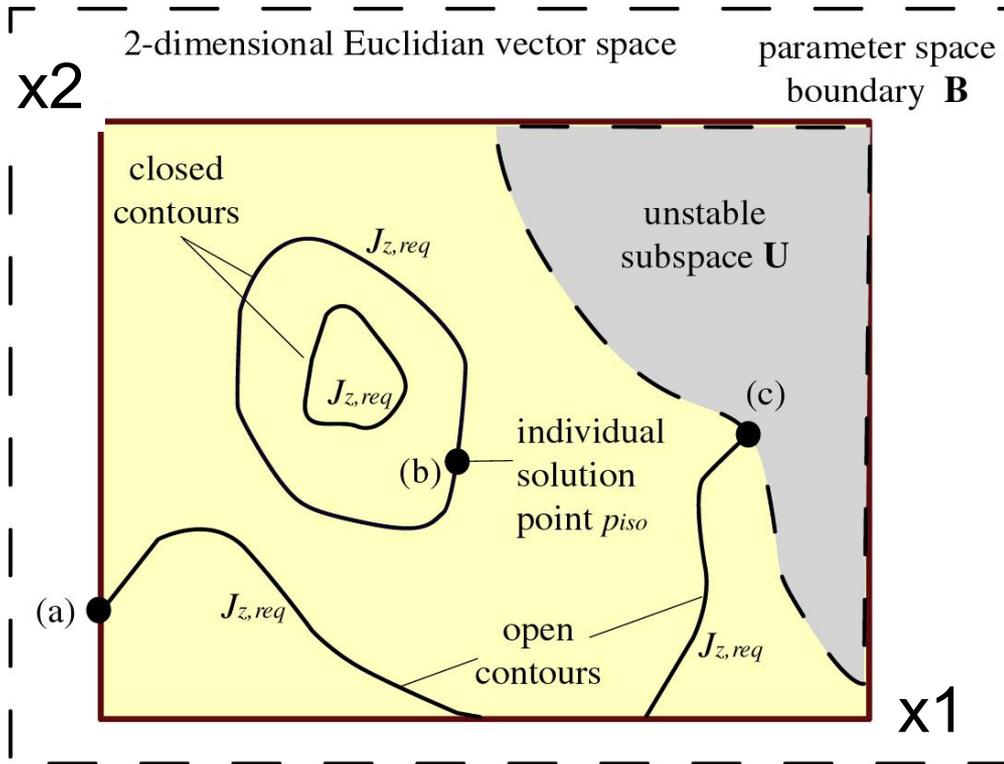
Performance $J_z(x_1, x_2): z = 1$

Variables $x_j, j = 1, 2: n = 2$

First Algorithm: **Exhaustive Search** coupled with bilinear interpolation

Number of points along j-th axis:

$$n_j = \left\lceil \frac{x_{j,UB} - x_{j,LB}}{\Delta x} \right\rceil$$



Can also use standard contouring code like MATLAB® `contourc.m`

k-th isoperformance point:

Taylor series expansion

$$J_z(x) = J_z(x^k) + \underbrace{\nabla J_z^T \Big|_{x^k} \Delta x}_{\text{first order term}} + \underbrace{\frac{1}{2} \Delta x^T H \Big|_{x^k} \Delta x}_{\text{second order term}} + \text{H.O.T.}$$

$$\nabla J_z = \begin{bmatrix} \frac{\partial J_z}{\partial x_1} \\ \frac{\partial J_z}{\partial x_2} \end{bmatrix}$$

$$\nabla J_z^T \Big|_{p^k} \Delta x \equiv 0$$

$$t^k = \mathfrak{R} \cdot \frac{-\nabla J_z \Big|_{p^k}}{\|\nabla J_z \Big|_{p^k}\|} = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \cdot n^k$$

t^k : tangential
step direction

$$\alpha_k = \left[2 \frac{\tau J_{z,req}}{100} t_k^T H \Big|_{x^k} t_k \right]^{-1/2}$$

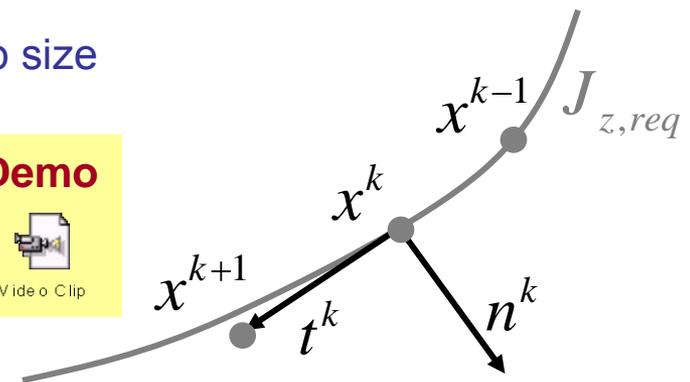
H: Hessian

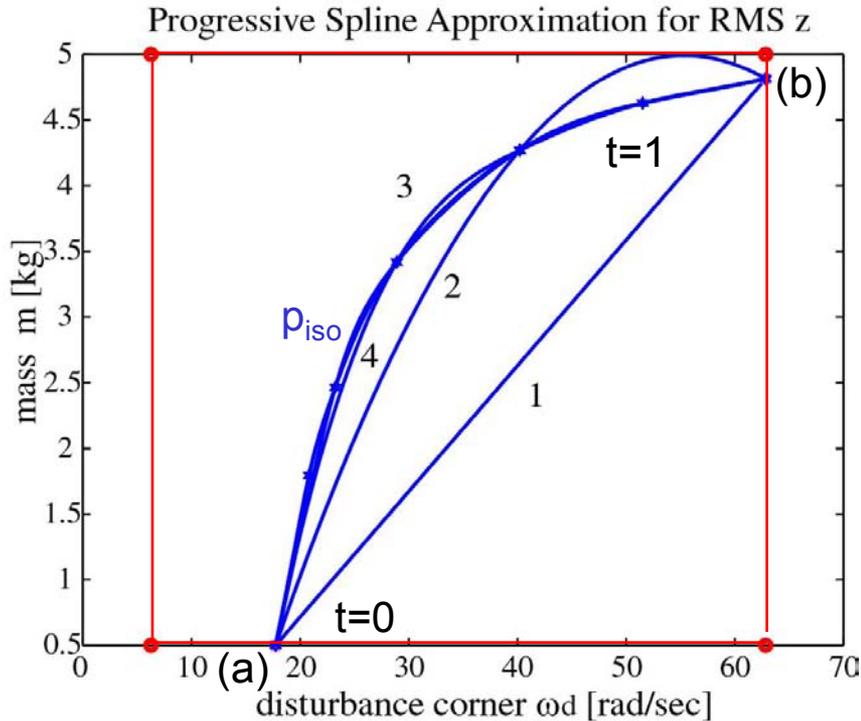
α^k : Step size

$$\Delta x = \alpha^k \cdot t^k$$

k+1-th isoperformance point:

$$x^{k+1} = x^k + \Delta x$$





- First find iso-points on boundary
- Then progressive spline approximation via segment-wise bisection
- Makes use of MATLAB spline toolbox , e.g. function `csape.m`

$$t \mapsto P_l t = \begin{bmatrix} x_{iso,1} & t \\ x_{iso,2} & t \end{bmatrix} = \begin{bmatrix} f_1 & t \\ f_2 & t \end{bmatrix}$$

$$t \in 0,1 \mapsto P_l t \in a,b$$

Use cubic
splines: $k=4$

$$f_{j,l} t = \sum_{i=1}^k \frac{t - \zeta_l}{k - i !} c_{j,l,k}^{k-i}, \quad t \in \zeta_l \dots \zeta_{l+1}$$

Bivariate Algorithm Comparison

Metric	Exhaustive Search (I)	Contour Follow (II)	Spline Approx (III)
FLOPS	2,140,897	783,761	377,196
CPU time [sec]	1.15	0.55	0.33
Tolerance τ	1.0%	1.0%	1.0%
Actual Error γ_{iso}	0.057%	0.347%	0.087%
# of isopoints	35	45	7

Results for SDOF Problem

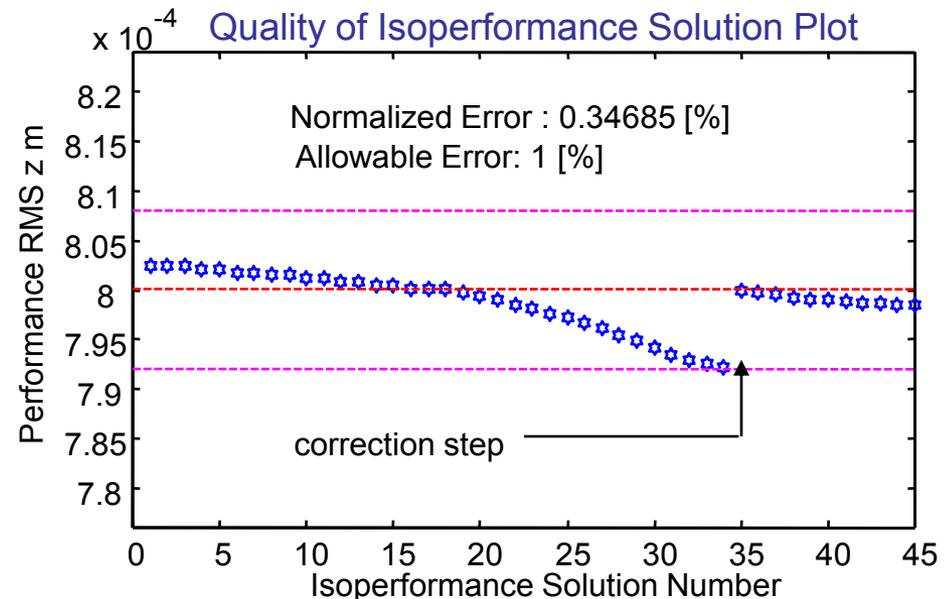
Conclusions:

- (I) most general but expensive
- (II) robust, but requires guesses
- (III) most efficient, but requires monotonic performance J_z

Isoperformance Quality Metric

“Normalized Error”

$$\gamma_{iso} = \frac{100}{J_{z,req}} \left[\frac{\sum_{r=1}^{n_{iso}} J_z x_{iso,k} - J_{z,req}}{n_{iso}} \right]^2 \Bigg|^{1/2}$$



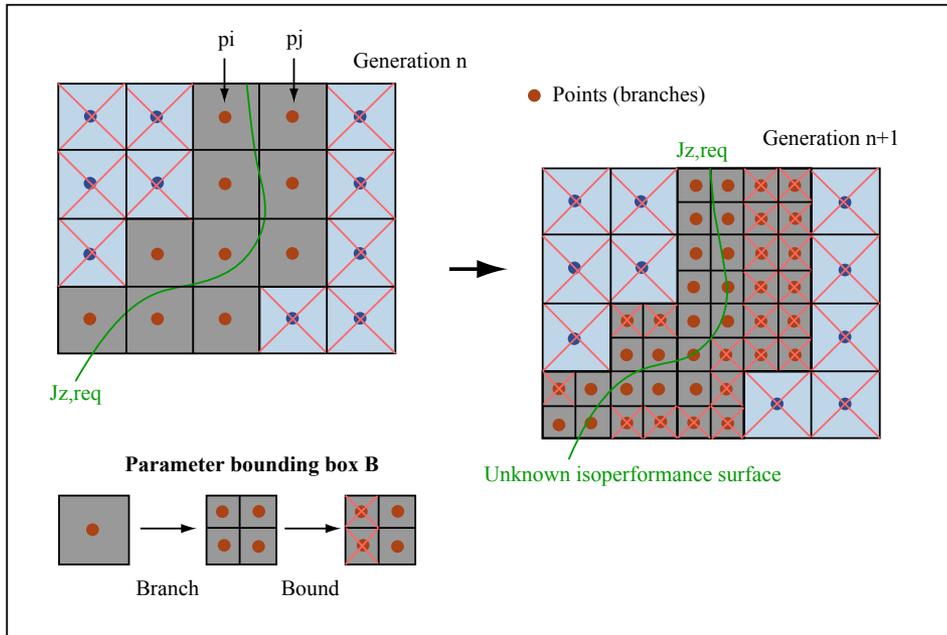
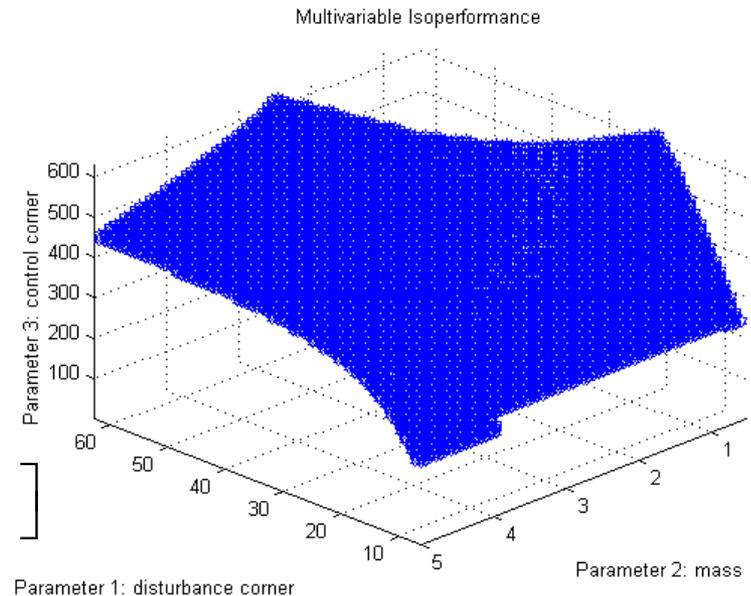


Image by MIT OpenCourseWare.

Exhaustive Search requires n_p -nested loops \rightarrow NP-cost: e.g.

$$N = \prod_{j=1}^{n_p} \left\lceil \frac{x_{UB,j} - x_{LB,j}}{\Delta x_j} \right\rceil$$



Branch-and-Bound only retains points/branches which meet the condition:

$$\left[J_z x_i \geq J_{z,req} \geq J_z x_j \right] \cup \left[J_z x_i \leq J_{z,req} \leq J_z x_j \right]$$

Expensive for small tolerance τ
Need initial branches to be fine enough

$$U = \begin{bmatrix} u_1 & \cdots & u_{n_z} \end{bmatrix}$$

$$\Sigma = \begin{bmatrix} \text{diag } \sigma_1 \cdots \sigma_{n_z} & 0_{n_z \times (n_p - n_z)} \end{bmatrix}$$

$$V = \begin{bmatrix} v_1 & \cdots & v_z & v_{z+1} & \cdots & v_n \end{bmatrix}$$

column space
nullspace

$$U \Sigma V^T = \nabla J_z^T$$

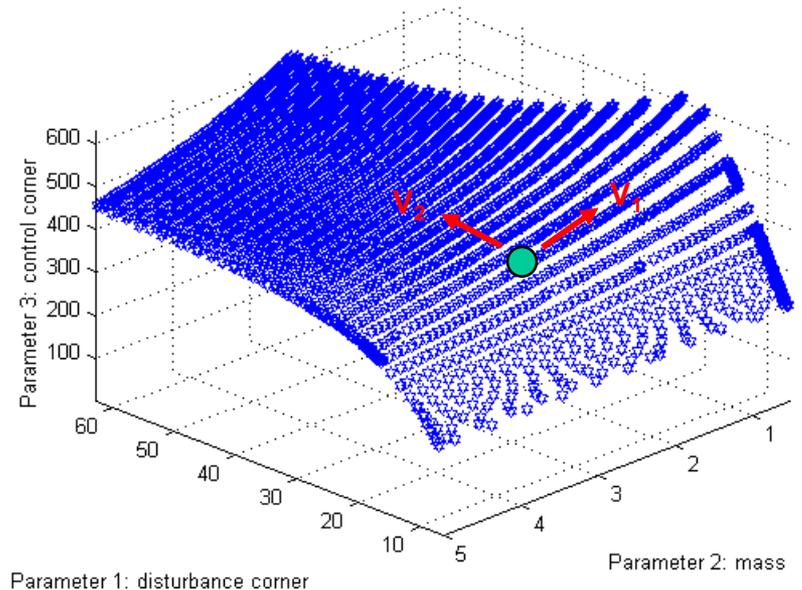
$$\nabla J_z = \begin{bmatrix} \frac{\partial J_1}{\partial x_1} & \cdots & \frac{\partial J_z}{\partial x_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial J_1}{\partial x_n} & \cdots & \frac{\partial J_z}{\partial x_n} \end{bmatrix}$$

Multivariable Isoperformance

$$\Delta x = \alpha \cdot \beta_1 v_{z+1} + \cdots + \beta_{n-z} v_n = \alpha V_t \beta$$

SVD of Jacobian provides V-matrix
V-matrix contains the orthonormal
vectors of the nullspace.

**Isoperformance set I is obtained by
following the nullspace of the Jacobian !**



Tangential front following is more efficient than branch-and-bound but can still be expensive for n_p large.

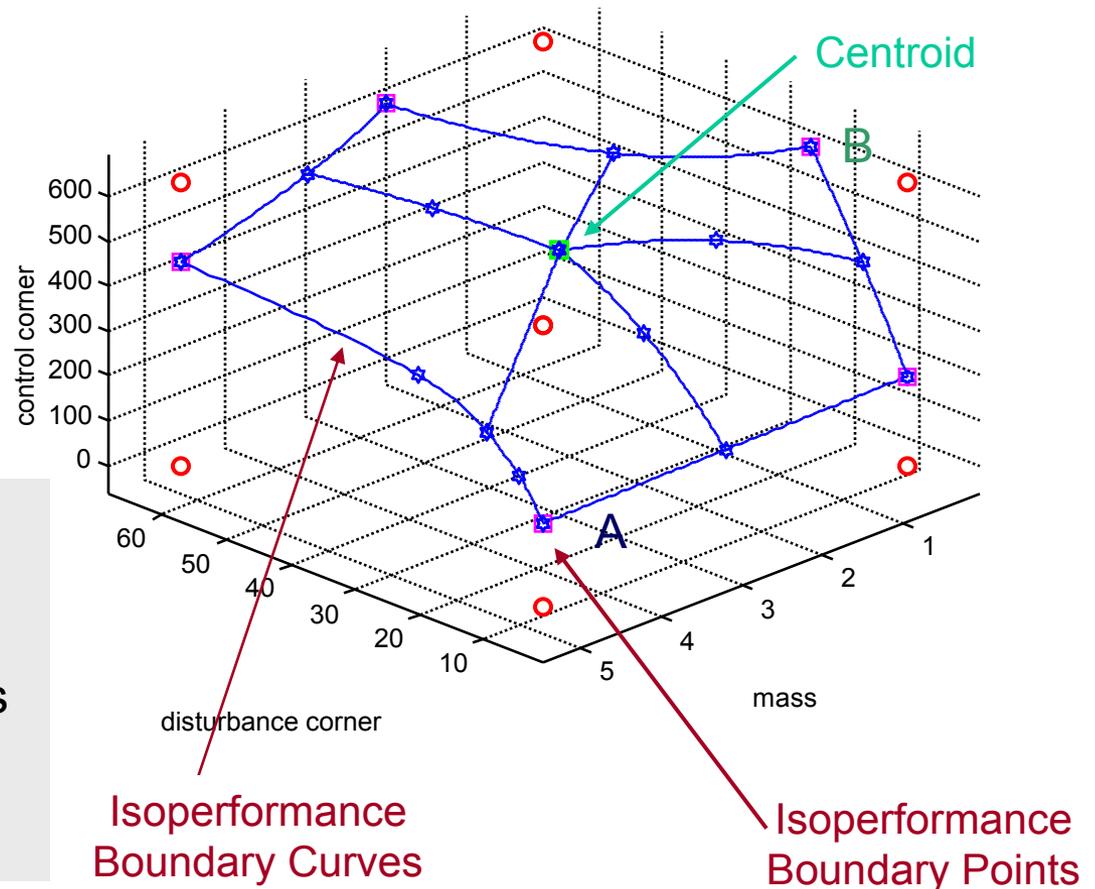
Idea: Find a representative subset off all isoperformance points, which are different from other.

“Frame-but-not-panels” analogy in construction

Algorithm:

1. Find Boundary (Edge) Points
2. Approximate Boundary curves
3. Find Centroid point
4. Approximate Internal curves

Vector Spline Approximation of Isoperformance Set



Challenges if $n_p > 2$

- Computational complexity as a function of $[n_z, n_d, n_p, n_s]$
- Visualization of isoperformance set in n_p -space

Problem Size:

$Z = \#$ of
performances

$d = \#$ of
disturbances

$n = \#$ of
variables

$n_s = \#$ of
states

Table: Multivariable Algorithm Comparison for SDOF ($n_p=3$)

Metric	Exhaustive Search	Branch-and-Bound	Tang Front Following	V- Spline Approx
MFLOPS	6,163.72	891.35	106.04	1.49
CPU [sec]	5078.19	498.56	69.59	4.45
Error Y_{iso}	0.87 %	2.43%	0.22%	0.42%
# of points	2073	7421	4999	20

From Complexity Theory: Asymptotic Cost

[FLOPS]

Exhaustive Search: $\log J_{exs} \rightarrow n_p \log \alpha + 3 \log n_s + c$

Branch-and-Bound: $\log J_{bab} \rightarrow n_g n_p \log 2 + \log \beta + 3 \log n_s + c$

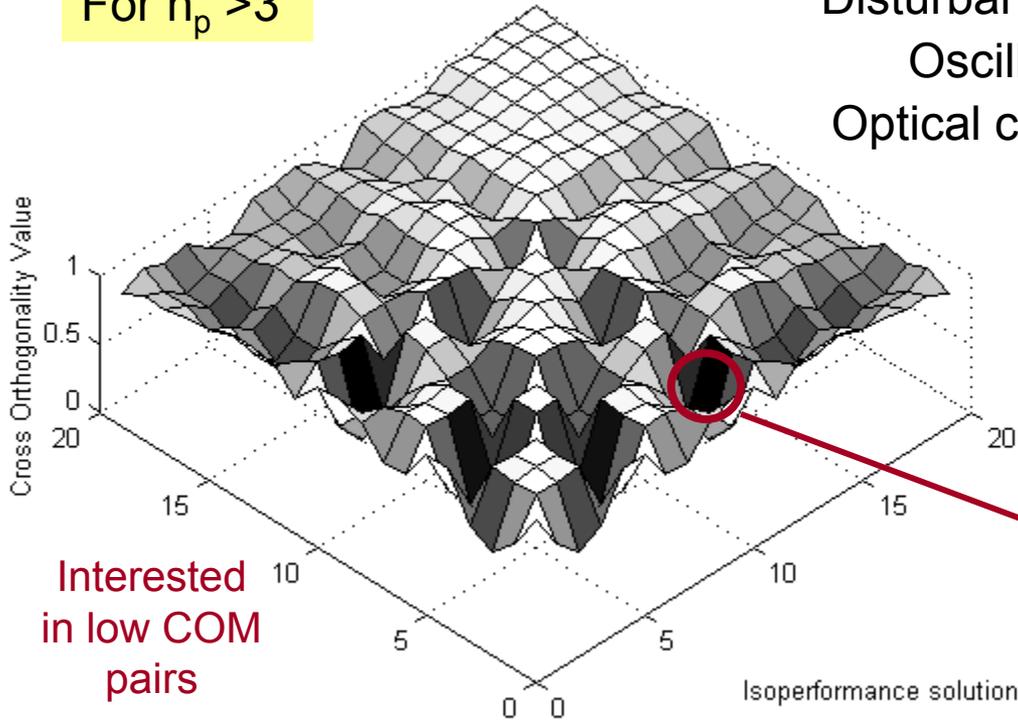
Tang Front Follow: $\log J_{tff} \rightarrow n_p - n_z \log \gamma + \log 1 + n_z + 3 \log n_s + c$

V-Spline Approx: $\log J_{vsa} \rightarrow n_p \log 2 + 3 \log n_s + \log(n_z + 1) + c$

Conclusion: Isoperformance problem is non-polynomial in n_p

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Engineering Systems Division and Dept. of Aeronautics and Astronautics

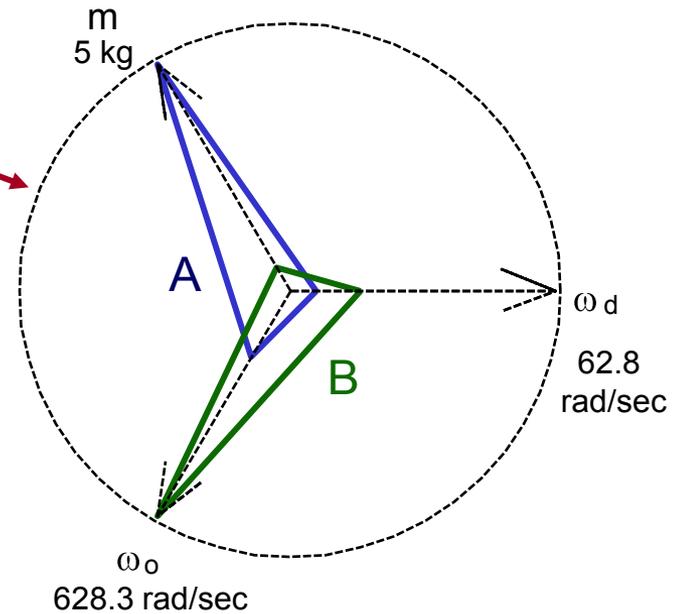
For $n_p > 3$



Disturbance corner ω_d	6.2832	21.3705
Oscillator mass m	5.0000	0.5000
Optical control bw ω_o	186.5751	628.3185
	A	B

Interested in low COM pairs

Multi-Dimensional Comparison of Isoperformance Points



Cross Orthogonality Matrix

$$COM(i, j) = \frac{P_{iso,i} \cdot P_{iso,j}}{|P_{iso,i}| \cdot |P_{iso,j}|}$$

Purpose of this case study:

Demonstrate the usefulness of Isoperformance on a realistic conceptual design model of a high-performance spacecraft

The following results are shown:

- Integrated Modeling
- Nexus Block Diagram
- Baseline Performance Assessment
- Sensitivity Analysis
- Isoperformance Analysis (2)
- Multiobjective Optimization
- Error Budgeting

Details are contained in CH7

on-orbit
configuration

Pro/E models
© NASA GSFC

Nexus
Spacecraft
Concept

launch
configuration

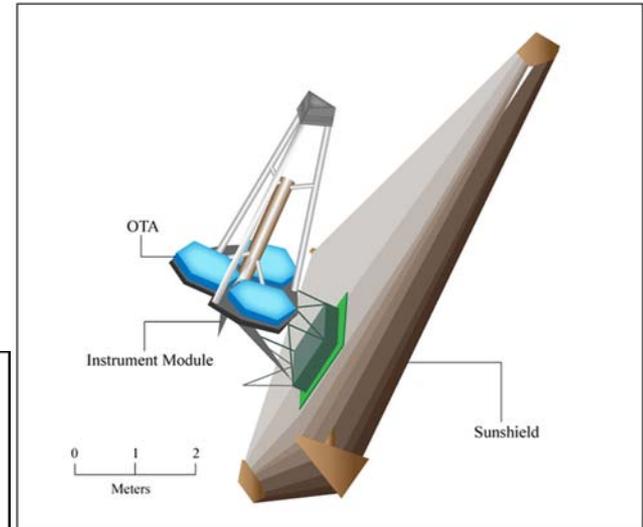


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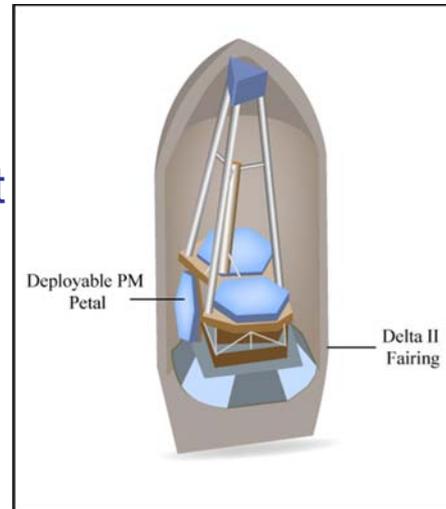
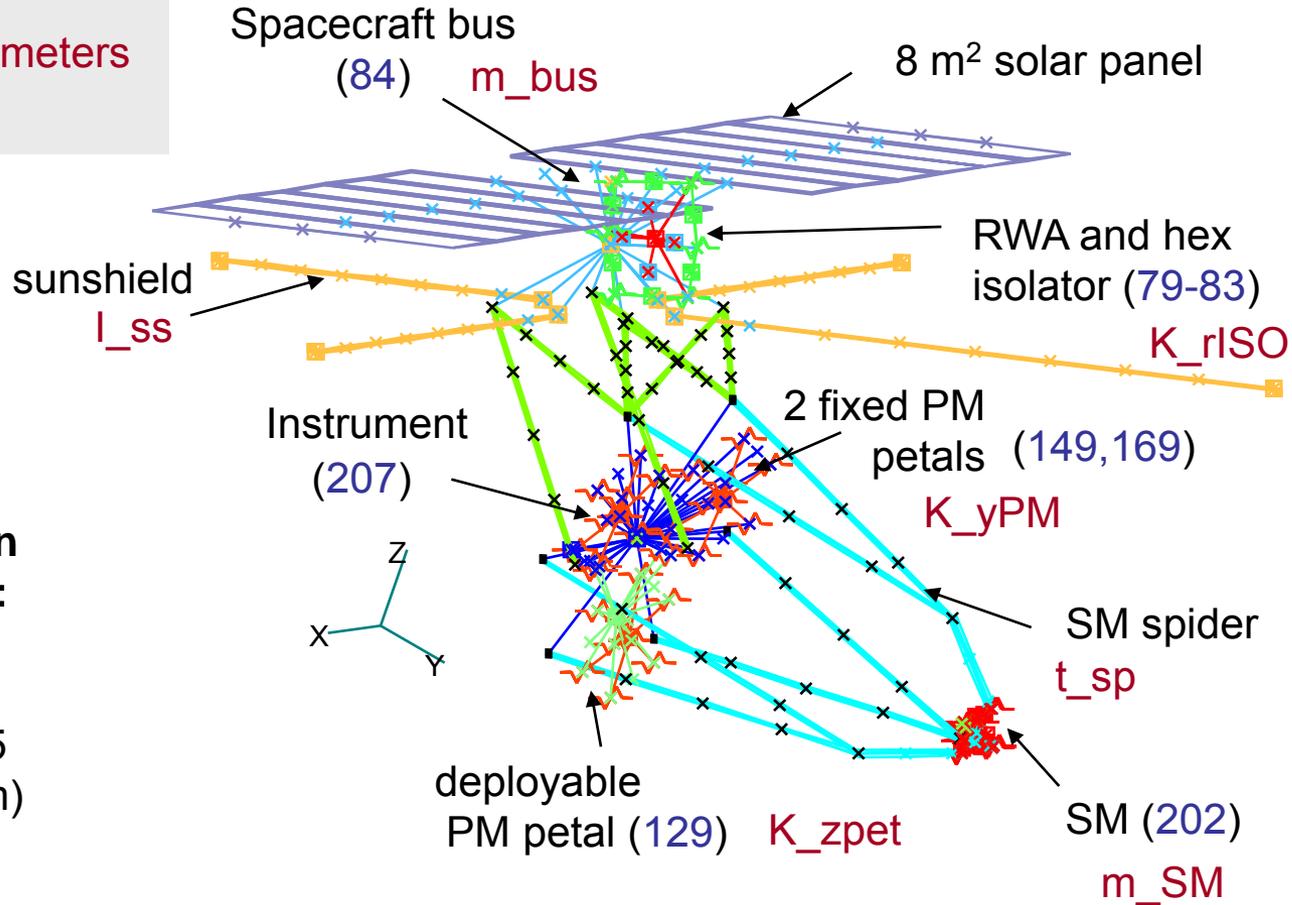


Image by MIT OpenCourseWare.

NGST Precursor Mission
2.8 m diameter aperture
Mass: 752.5 kg
Cost: 105.88 M\$ (FY00)
Target Orbit: L2 Sun/Earth
Projected Launch: 2004

Legend:

Design Parameters
(I/O Nodes)



Cassegrain Telescope:

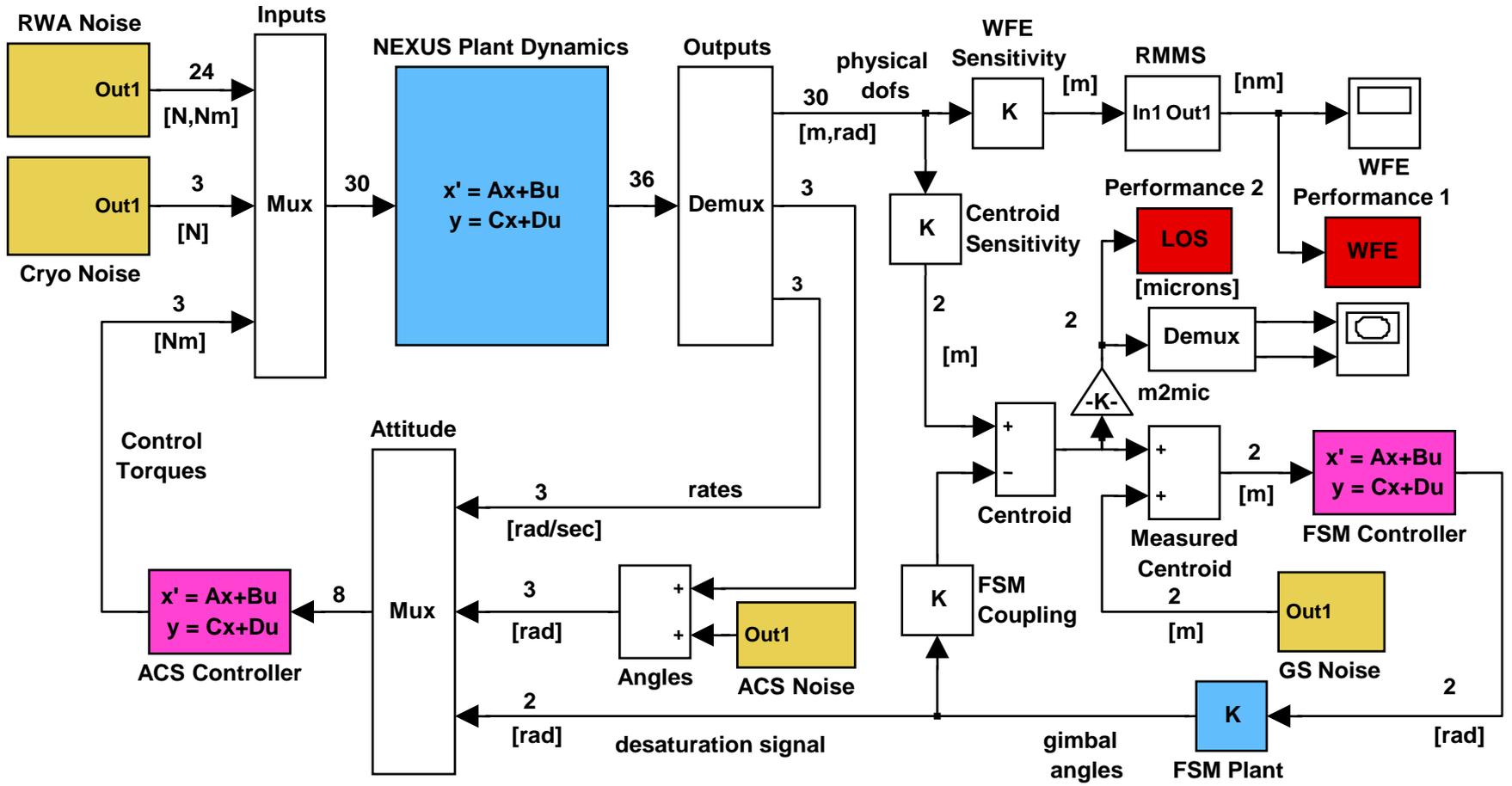
PM (2.8 m)
PM f/# 1.25
SM (0.27 m)
f/24 OTA

**Structural Model (FEM)
(Nastran, IMOS)**

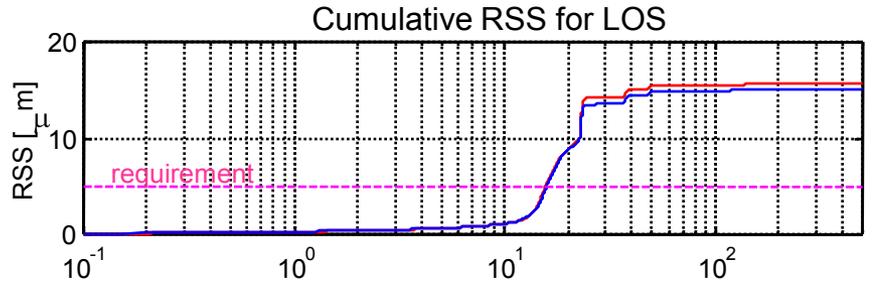
→ Ω, Φ

Number of performances: $n_z=2$
 Number of design parameters: $n_p=25$

Number of states $n_s=320$
 Number of disturbance sources: $n_d=4$

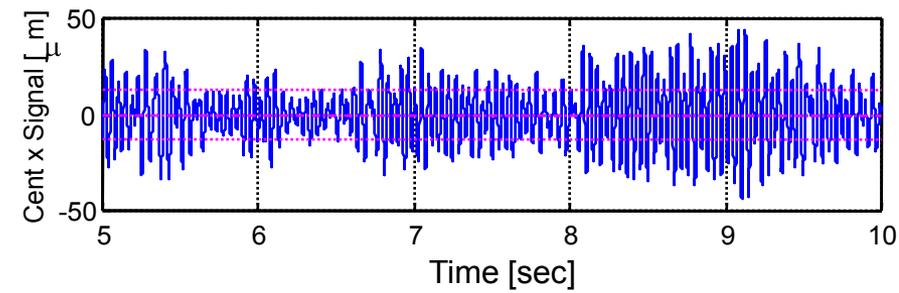
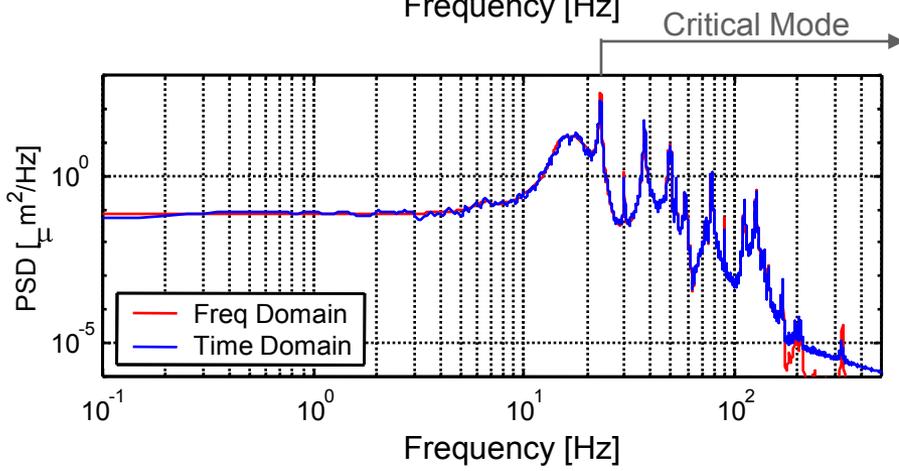


$J_z(p^\circ)$

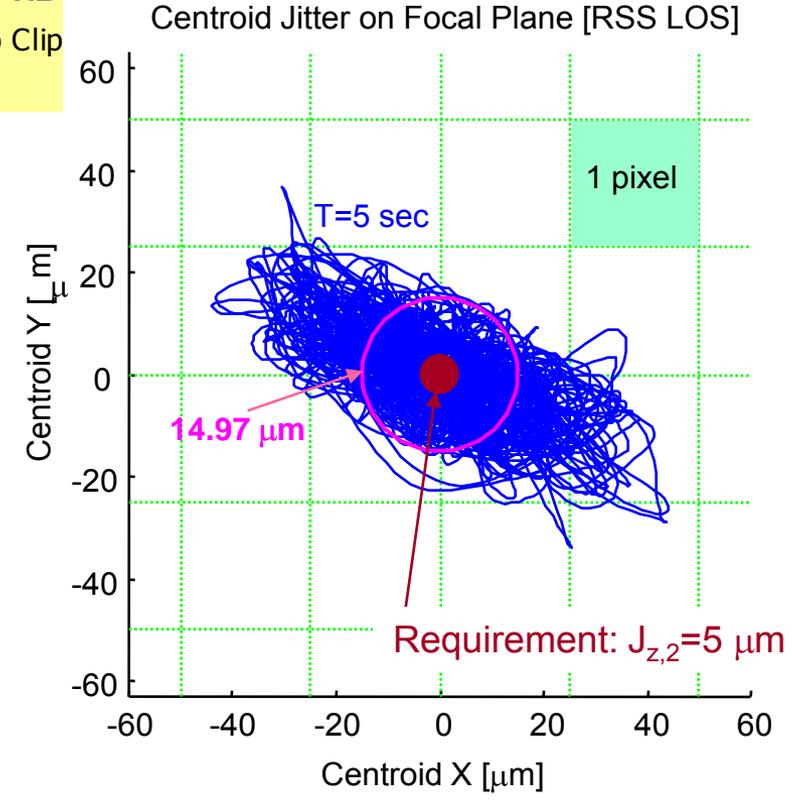


Results

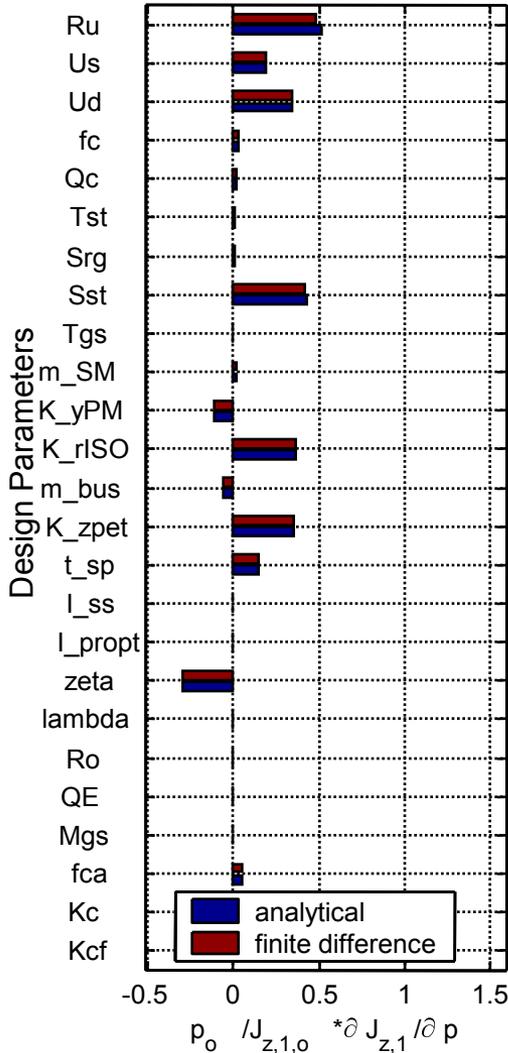
	Lyap/Freq	Time	
$J_{z,1}$ (RMMS WFE)	25.61	19.51	[nm]
$J_{z,2}$ (RSS LOS)	15.51	14.97	[μm]



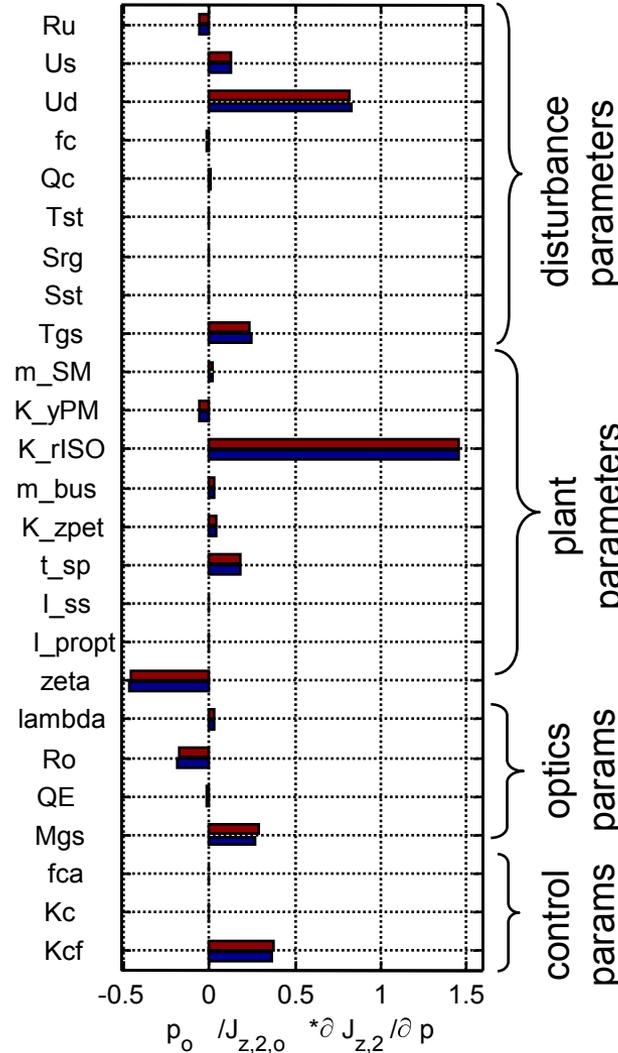
23.1 Hz
Video Clip



Norm Sensitivities: RMMS WFE



Norm Sensitivities: RSS LOS



Graphical Representation of Jacobian evaluated at design p_o , normalized for comparison.

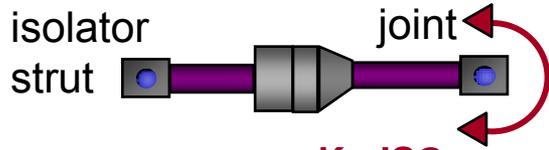
$$\bar{\nabla} J_z = \frac{p_o}{J_{z,o}} \begin{bmatrix} \frac{\partial J_{z,1}}{\partial R_u} & \frac{\partial J_{z,2}}{\partial R_u} \\ \dots & \dots \\ \frac{\partial J_{z,1}}{\partial K_{cf}} & \frac{\partial J_{z,2}}{\partial K_{cf}} \end{bmatrix}$$

RMMS WFE most sensitive to:

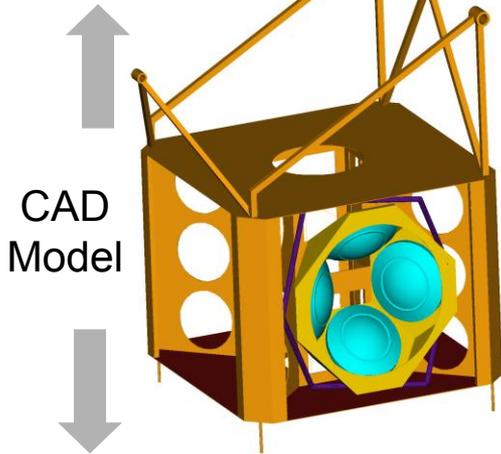
- Ru - upper op wheel speed [RPM]
- Sst - star track noise 1σ [asec]
- K_rISO - isolator joint stiffness [Nm/rad]
- K_zpet - deploy petal stiffness [N/m]

RSS LOS most sensitive to:

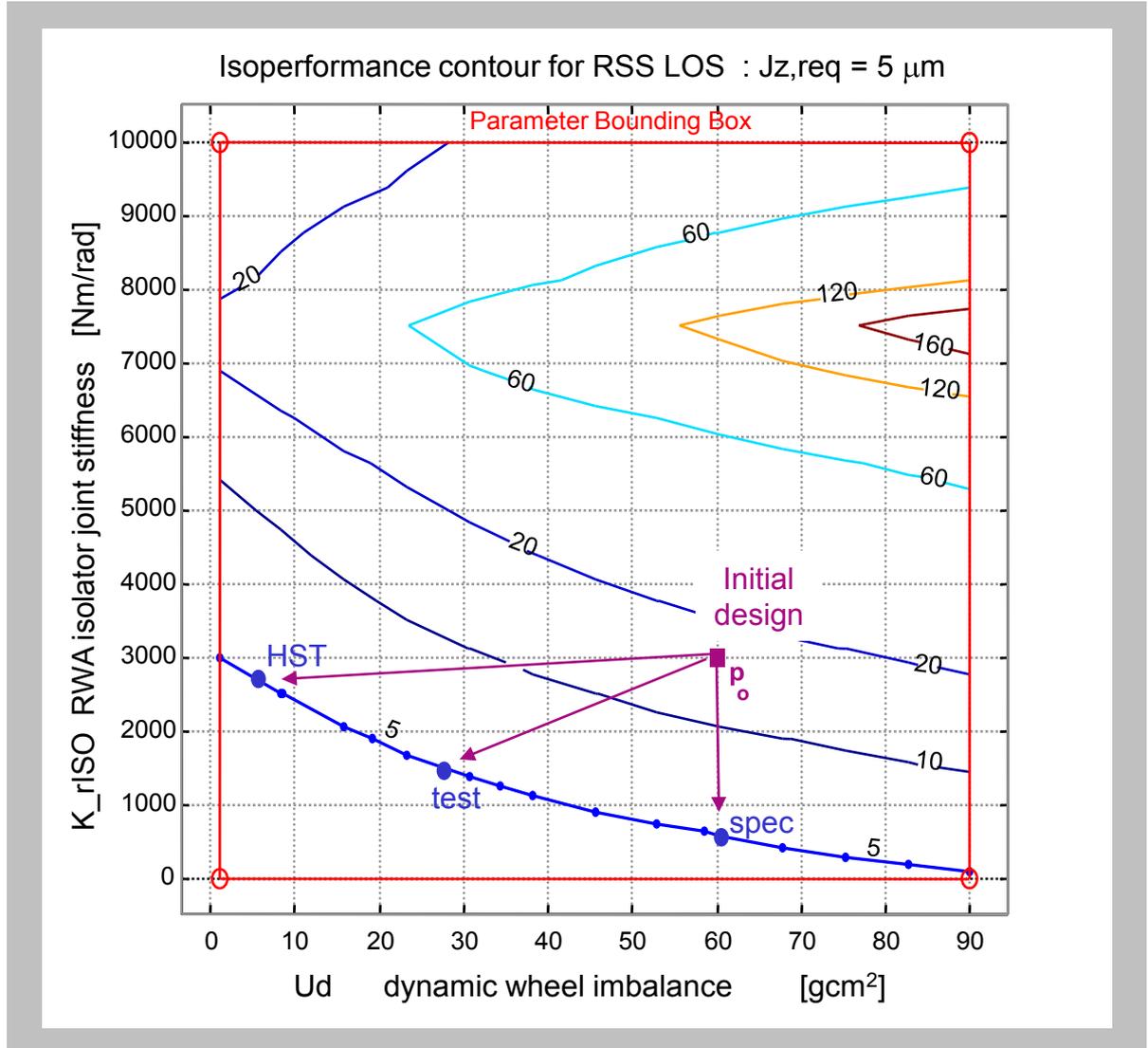
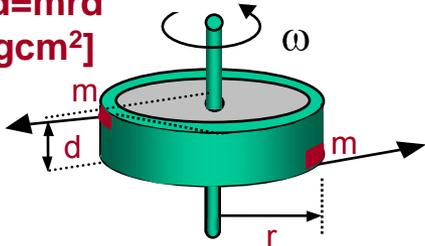
- Ud - dynamic wheel imbalance [gcm²]
- K_rISO - isolator joint stiffness [Nm/rad]
- zeta - proportional damping ratio [-]
- Mgs - guide star magnitude [mag]
- Kcf - FSM controller gain [-]



K_{rISO}
[Nm/rad]



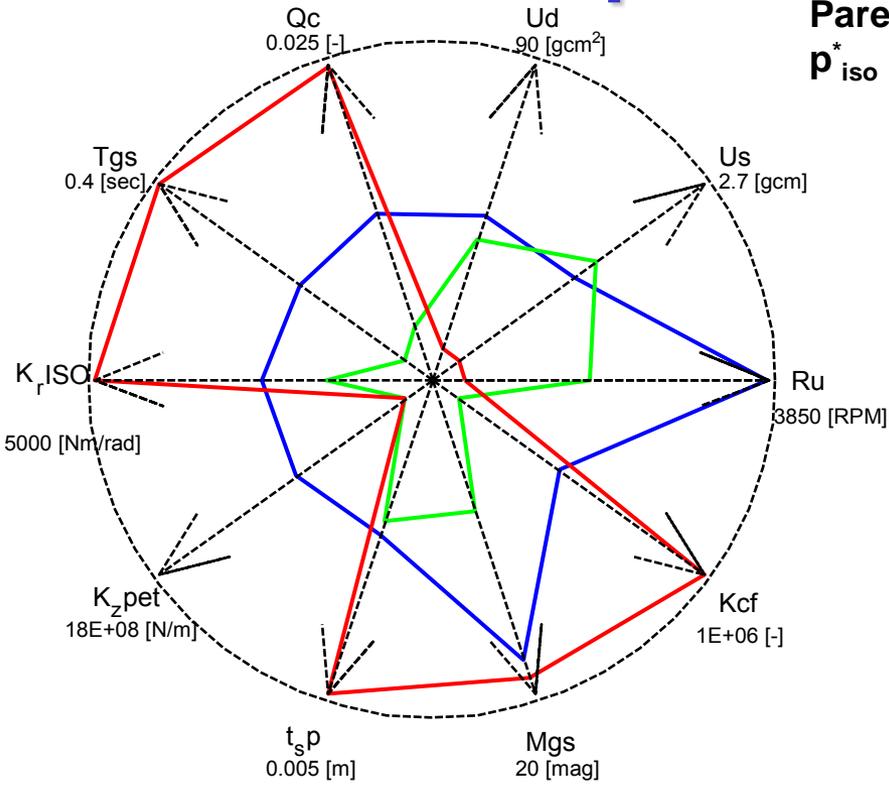
$Ud = mrd$
[gcm²]



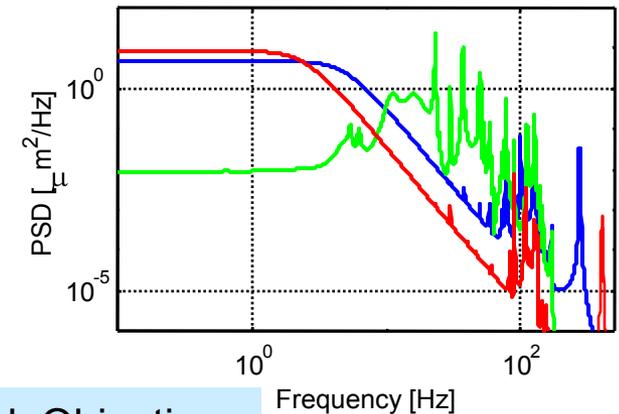
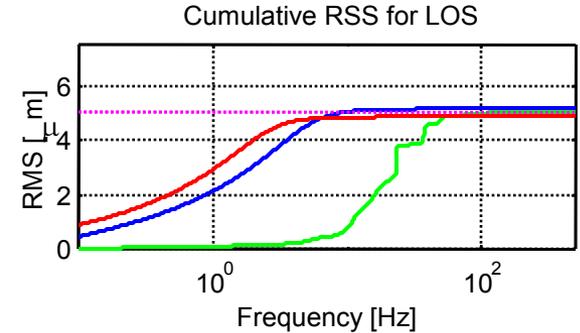
Isoperformance $n_p = 10$

Pareto-Optimal Designs

p^*_{iso}



- Design A**
Best "mid-range" compromise
- Design B**
Smallest FSM control gain
- Design C**
Smallest performance uncertainty



Performance

Cost and Risk Objectives

- A: $\min(J_{c1})$
- B: $\min(J_{c2})$
- C: $\min(J_{r1})$

	Jz,1	Jz,2	Jc,1	Jc,2	Jr,1
Design A	20.0000	5.2013	0.6324	0.4668	+/- 14.3218 %
Design B	20.0012	5.0253	0.8960	0.0017	+/- 8.7883 %
Design C	20.0001	4.8559	1.5627	1.0000	+/- 5.3067 %

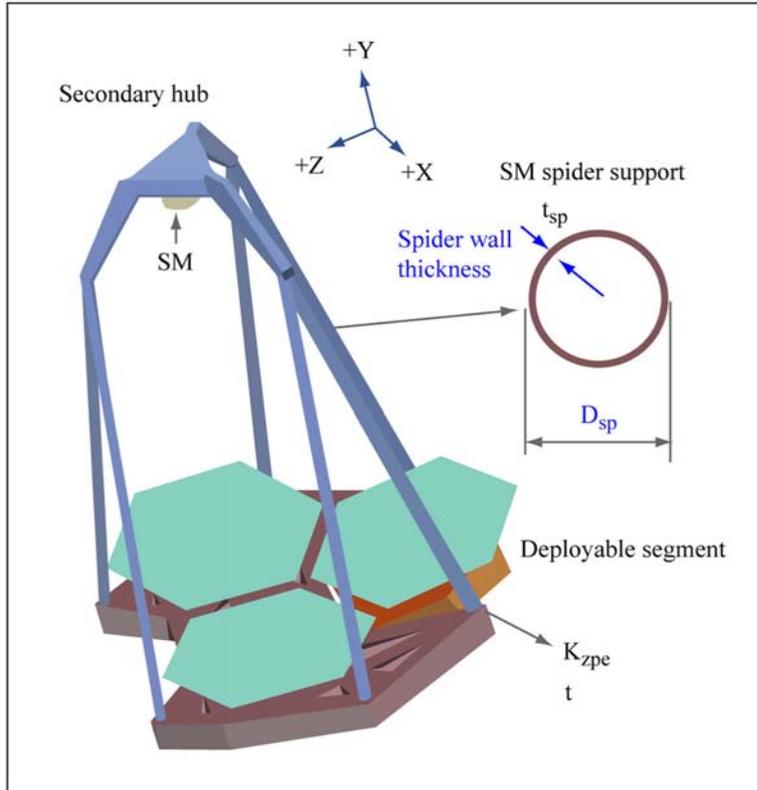
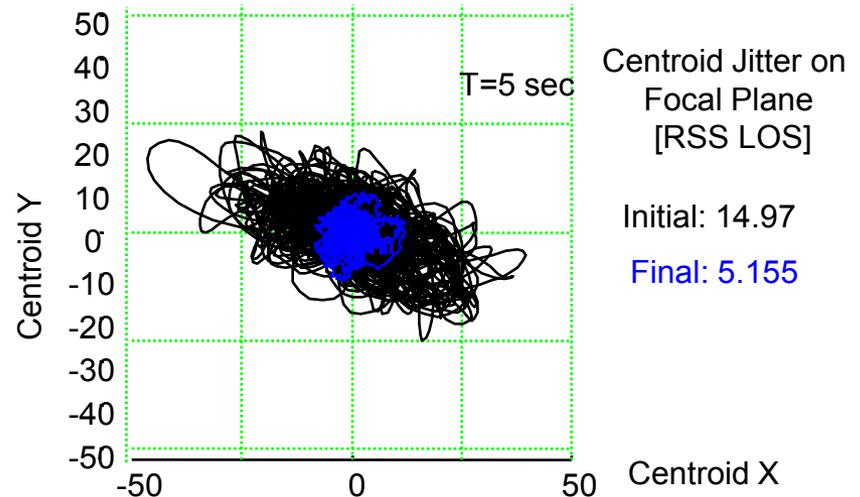


Image by MIT OpenCourseWare.

Improvements are achieved by a well balanced mix of changes in the disturbance parameters, structural redesign and increase in control gain of the FSM fine pointing loop.

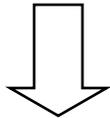
Parameters	Initial	Final	
Ru	3000	3845	[RPM]
Us	1.8	1.45	[gcm]
Ud	60	47.2	[gcm ²]
Qc	0.005	0.014	[-]
Tgs	0.040	0.196	[sec]
KrISO	3000	2546	[Nm/rad]
Kzpet	0.9E+8	8.9E+8	[N/m]
tsp	0.003	0.003	[m]
Mgs	15	18.6	[Mag]
Kcf	2E+3	4.7E+5	[-]



Example: Baseball season is starting soon !

What determines success of a team ?

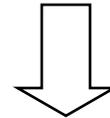
Pitching



ERA

“Earned Run Average”

Batting



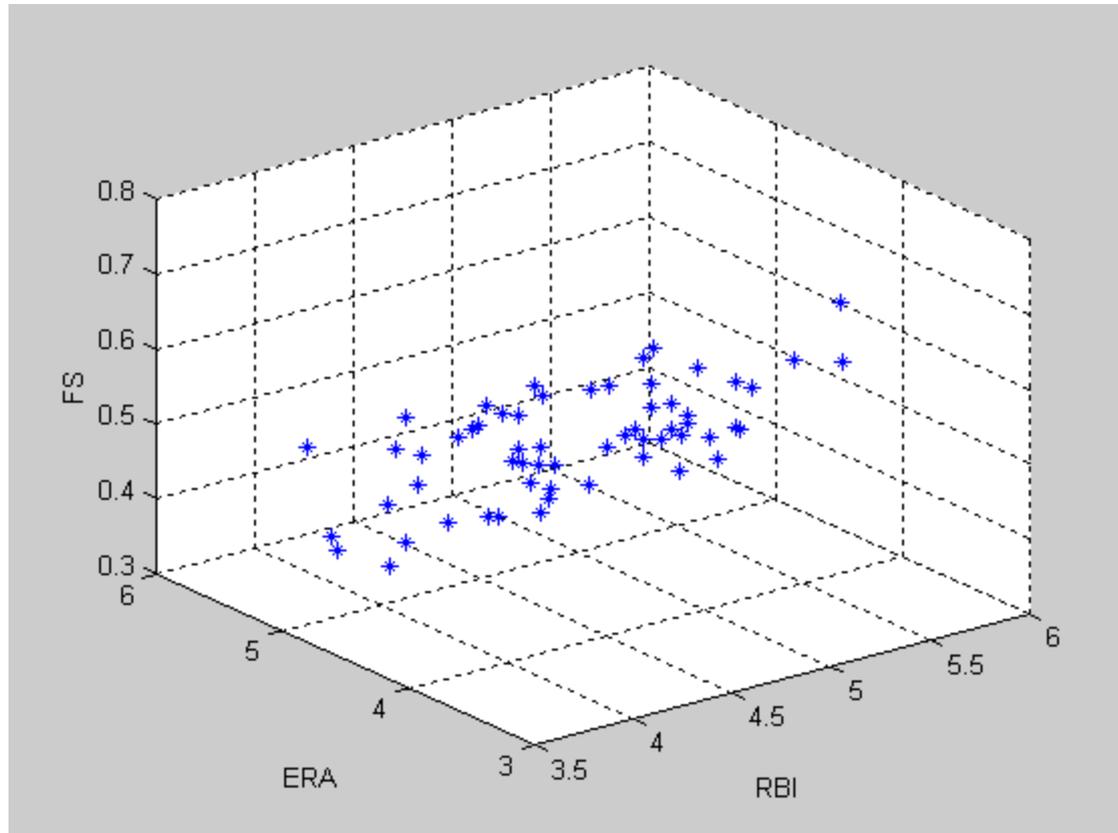
RBI

“Runs Batted In”

How is success of team measured ?

$FS = \text{Wins/Decisions}$

Team results for 2000, 2001 seasons: RBI,ERA,FS



Step-by-step process for obtaining (bivariate) isoperformance curves given statistical data:

Starting point, need:

- Model - derived from empirical data set
- (Performance) Criterion
- Desired Confidence Level

Step 1: Obtain an expression from model for expected performance of a “system” for individual design i as a function of design variables $x_{1,i}$ and $x_{2,i}$

1.1 assumed model

$$E J_i = a_0 + a_1 x_{1,i} + a_2 x_{2,i} + a_{12} (x_{1,i} - \bar{x}_1)(x_{2,i} - \bar{x}_2) \quad (1)$$

1.2 model fitting

mean

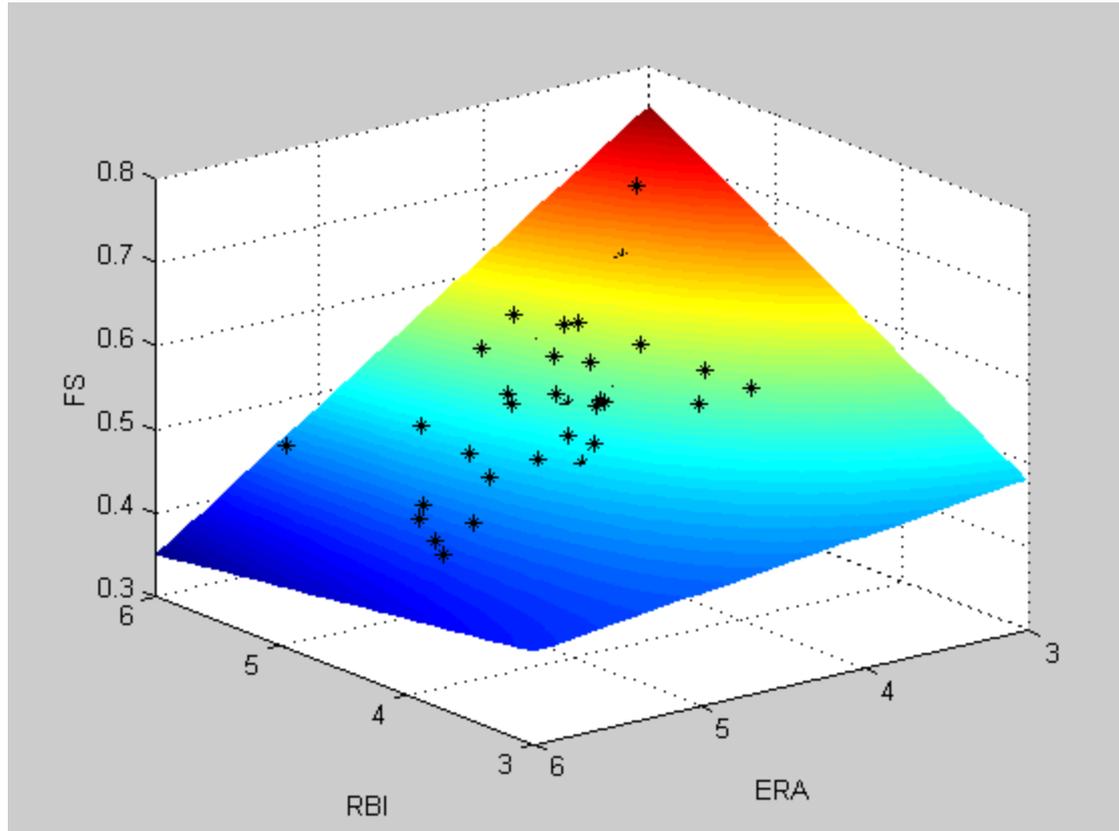
$$a_0 = \frac{1}{N} \sum_{j=1}^N J_j$$

E.g. use MATLAB
fminunc.m for
optimal surface fit

Baseball:

Obtain an expression for expected final standings (FS_i) of individual Team i as a function of RBI_i and ERA_i

$$E FS_i = m + a RBI_i + b ERA_i + c (RBI_i - \overline{RBI})(ERA_i - \overline{ERA})$$



Coefficients:

$$a_0 = 0.7450$$

$$a_1 = 0.0321$$

$$a_2 = -0.0869$$

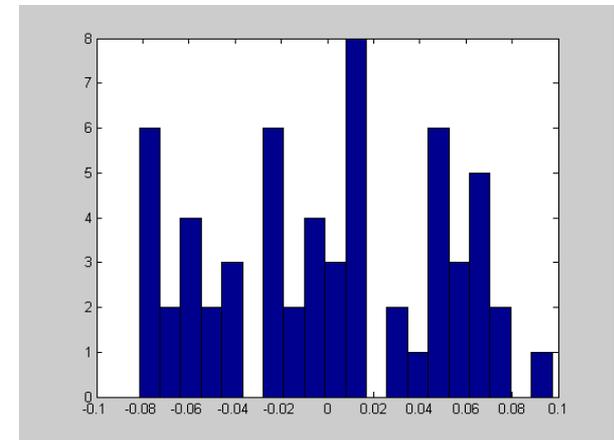
$$a_{12} = -0.0369$$

RMSE:

Error

$$\sigma_e = 0.0493$$

Error
Distribution

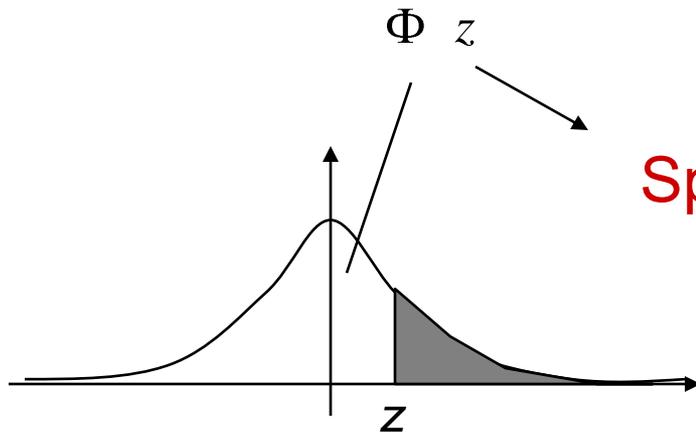


Step 2: Determine expected level of performance for design i such that the probability of adequate performance is equal to specified confidence level

$$E J_i = J_{req} + z\sigma_\varepsilon \quad (2)$$

Required
performance
level

Error Term
(total variance)



Specify

Confidence level
normal variable z
(Lookup Table)

$$\Phi z = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{z^2}{2}} dz$$

Baseball:

Performance criterion

- User specifies a final desired standing of $FS_i=0.550$

Confidence Level

- User specifies a .80 confidence level that this is achieved

Spec is met if for Team i :

$$E FS_i = .550 + z\sigma_r = .550 + 0.84 \cdot 0.0493 = 0.5914$$

From normal table lookup

Error term from data

If the final standing of team i is to equal or exceed .550 with a probability of .80, then the expected final standing for Team i must equal 0.5914

Step 3: Put equations (1) and (2) together

$$J_{req} + z\sigma_r = E J_i = a_0 + a_1 x_{1,i} + a_2 x_{2,i} + a_{12} (x_{1,i} - \bar{x}_1)(x_{2,i} - \bar{x}_2) \quad (3)$$

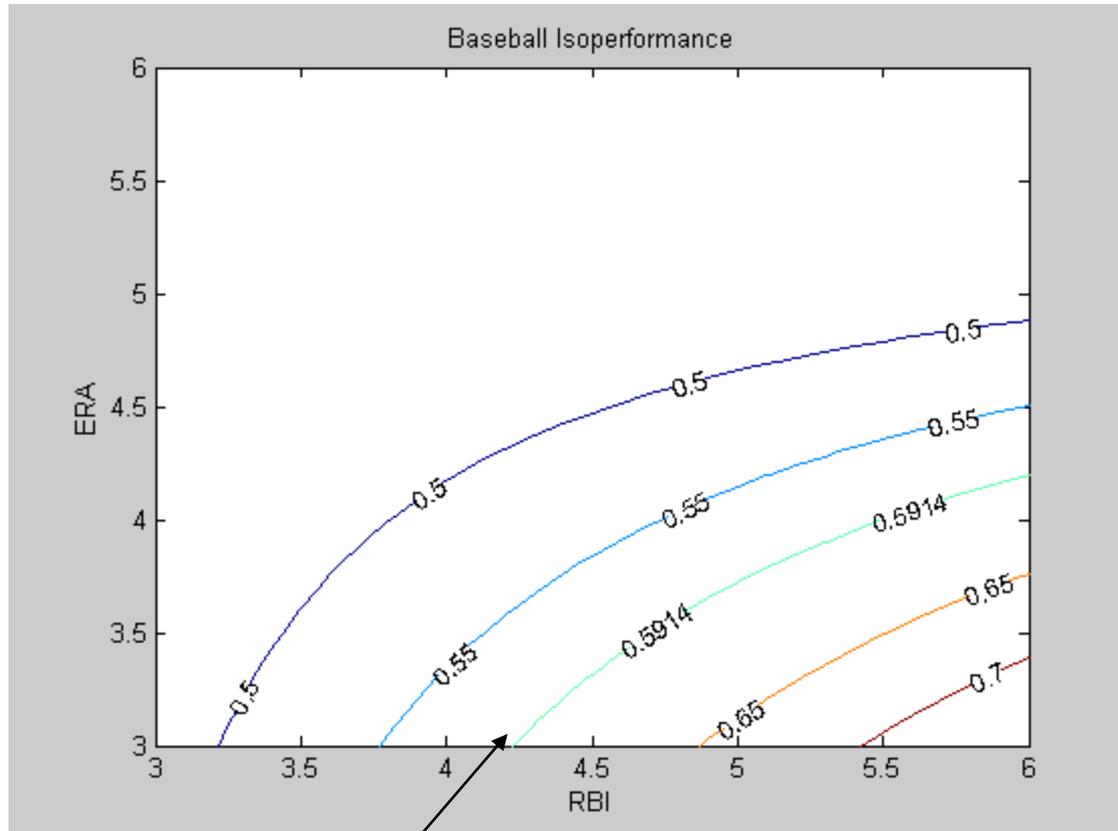
- Four constant parameters: a_0, a_1, a_2, a_{12}
- Two sample statistics: \bar{x}_1, \bar{x}_2
- Two design variables: $x_{1,i}$ and $x_{2,i}$

Then rearrange: $x_{2,i} = f(x_{1,i})$

Baseball:

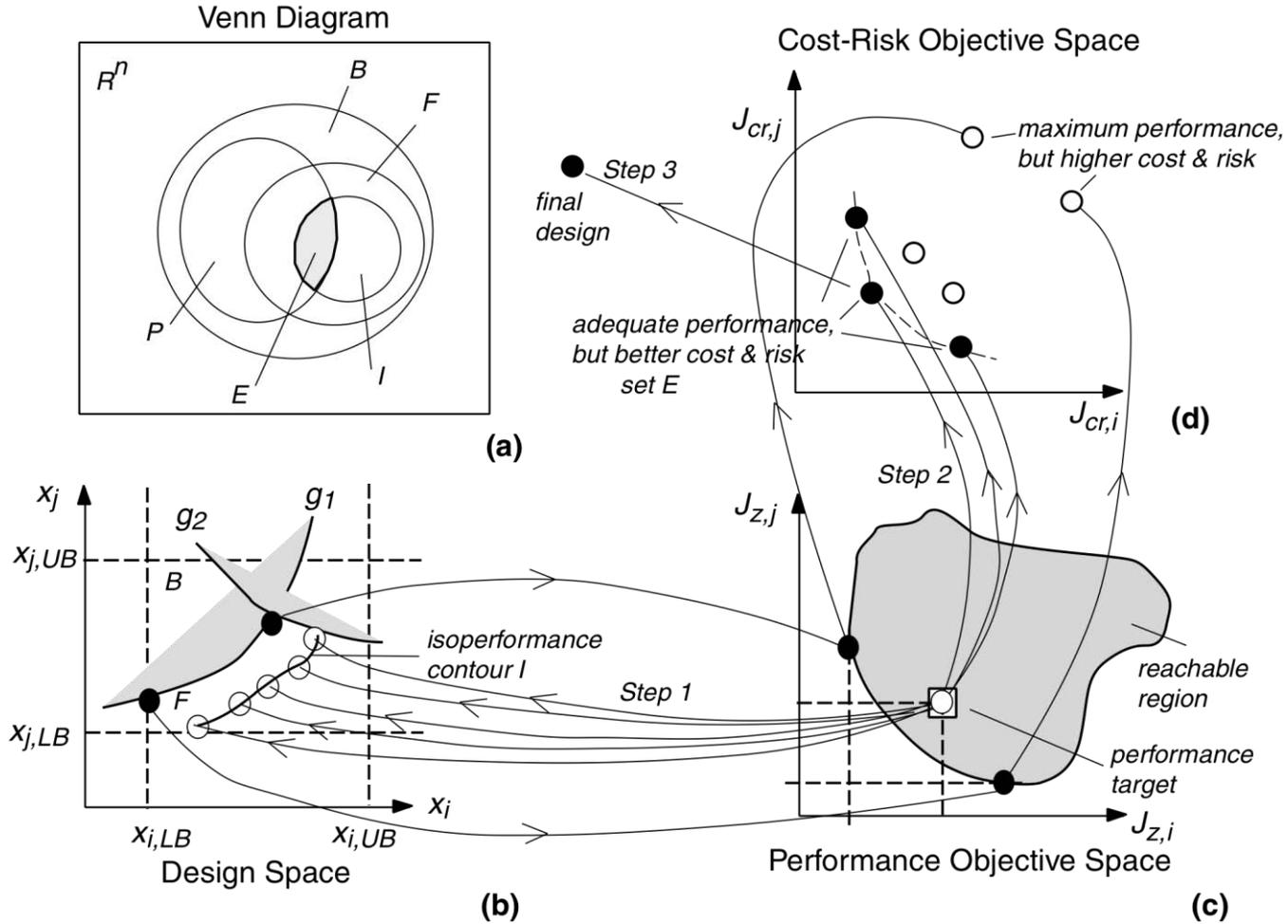
$$RBI_i = \frac{.5914 - m - bERA_i + c\overline{RBI} \overline{ERA_i} - \overline{ERA}}{a + c \overline{ERA_i} - \overline{ERA}}$$

Equation
for isoperformance
curve



This is our desired tradeoff curve

- Traditional process goes from design space \mathbf{x} \rightarrow objective space \mathbf{J} (forward process)
- Many systems are designed to meet “targets”
 - Performance, Cost, Stability Margins, Mass ...
- Methodological Options
 - Formulate optimization problem with equality constraints given by targets
 - Goal Programming minimizes the “distance” between a desired “target” state and the achievable design
 - Isoperformance finds a set of (non-unique) performance invariant solutions \rightarrow multiple solutions
- Isoperformance works backwards from objective space \mathbf{J} \rightarrow design space \mathbf{x} (reverse process)
 - Deterministically
 - Stochastically



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de Weck, O.L. and Jones M. B., “Isoperformance: Analysis and Design of Complex Systems with Desired Outcomes”, *Systems Engineering*, 9 (1), 45-61, January 2006

de Weck O.L., Miller D.W., “Multivariable Isoperformance Methodology for Precision Opto-Mechanical System”, Paper AIAA-2002-1420, 43rd AIAA/ASME /ASCE/AHS Structures, Structural Dynamics, and Materials Conference, Denver, Colorado, April 22-25, 2002

Schniederjans MJ, *Goal programming Methodology and Applications*, Kluwer Publishers, Boston, 1995

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