## NASA Instrument Cost Model

## NICM

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## NICM Introduction

- NICM is the NASA Instrument Cost Model
- Parametric cost model for NASA's space flight instruments
- Operates at the Instrument System and Subsystem Levels
- Supports Remote Sensing and In-situ instruments
- NICM is used across all NASA centers and is also available to restricted release to external organizations.
- Built off 174 previously flow instruments



## Current NICM Dataset

- Collected data for 262 instruments
- Normalized database
- 174 of the 262 normalized
- 111 remote sensing instruments
- 49 in-situ instruments
- Remote Sensing Instruments Types:
- Optical, Active micro/sub-millimeter wave, Passive micro/sub-millimeter wave, Particles, and Fields
- In-situ Types based on instrument mounting:
- Body, Arm/Mast, Atmospheric Probe.


## Data Ground Rules \& Assumptions

- Includes only instruments launched 1985 and after
- Excludes $100 \%$ foreign built instruments
- However includes some foreign contributed subsystems
- Includes space flight remote sensing and in-situ instruments only
- Includes costs of development summed over phases $\mathrm{B}, \mathrm{C} \& \mathrm{D}$ (through Launch +30 days)
- Excludes advance studies, pre-phase A and phase A costs.
- Excludes advanced technology development costs - TRL 1, 2, 3
- Excludes costs for science teams, ground data development and mission operations.
- Includes only development of $1^{\text {st }}$ unit cost
- Excludes subsequent modified builds or copies
- Did not estimate nonrecurring or recurring cost


## Data Ground Rules \& Assumptions

- Database costs are expressed in FY04 \$K. The tools have the capability to express costs in any fiscal year's dollars using the NASA New Start Inflation Indices.
- Full cost accounting practice is assumed for all NASA centers.
- Cost data are assumed to include fee.


## NICM Dataset By Instrument Lead Organizations

Total Normalized Instruments: 160



## In-Situ Instrument Mounting Types

n2

## Total Normalize Instruments: <br> 49

- Examples
$\begin{array}{lc}\text { mples: } & \text { Probe } \\ \text { - Huygens Probe on } & 22 \% \\ \text { Cassini } & \\ \text { •Galileo's Probe } & \\ & \\ & \end{array}$



## NICM Tool Strengths

(ax)

- Based on high quality dataset
- Models validated by statistical analysis
- Reviewed by subject area experts
- Complete audit trail and documentation
- Provides probabilistic cost predictions
- Allows uncertainty for inputs
- Calculates S-curve for cost \& schedule
- Captures Objective Information
- No adjustable "knobs"
- User friendly database search engine
- Searches the normalized database for analogy instruments
- Provides Joint Confidence Level (JCL) Analysis


## Model Limitations

- NICM VI costing tool does not estimate the following:
- Airborne instruments
- Suites of instruments
- Specialty subsystems, e.g. engineering experiments or demonstrations (e.g. Electra on MRO).
- Advanced technology developments
- Nonrecurring or recurring costs
- Copies/multiple builds
- Resource estimates, e.g. labor, materials, services, etc.


## Methodology

- Cluster Analysis
- Identifies Instrument Groupings from Attribute Values
- Assesses Consistency of Groups with Instrument Types
- Principal Components Analysis
- Standard Data Mining Technique that
- Finds Significant Cost Drivers from Instrument Attributes
- Identifies Instrument Data Outliers - Revisit data with technical experts
- Bootstrap Cross Validation
- Bootstrap: Process for generating meaningful statistics without assuming asymptotic normality.
- Cross Validation: Partitioning of data set into training and testing sets. Out-of-sample validation.


## Cluster Analysis - Remote Sensing Instrument

Na


## Bootstrap Cross Validation

## Instrument

- Explanation of ".632"

Bootstrap Cross-validation

| \# | Trial \#1 | Trial \#2 | ... | Trial \#999 |
| :---: | :---: | :---: | :---: | :---: |
| 1 | II | II |  | $\Delta_{1009}$ |
| 2 | 1 | $\Delta_{2,2}$ |  | II |
| 3 | 1 | 1 |  | $\Delta_{3099}$ |
| 4 | $\Delta_{4,1}$ | 1 | .. | I |
| 5 | // | // |  | III |
| 6 | 1 | 1 |  | 1 |
| 7 | $\Delta_{71}$ | $\Delta_{72}$ |  | $\Delta_{7,999}$ |
| 8 | I | 1 | $\ldots$ | 1 |
| 9 | /I | 1 |  | II |
| 10 | $\Delta_{10,1}$ | 1 | $\ldots$ | $\Delta_{10,999}$ |

for ach CER (\& associated for each CER (\& associated dataset)

- Sample with replacement from the dataset (using sample size same as dataset)
- Fit regression model to trial sample selection
- Predict cost with model for instruments in original dataset 9 that were not selected by trial 10 sampling for testing
- Repeat above steps 999 times, saving cost deltas for each instrument tested
- Calculate average model variance ( $=$ cost delta^2) for all 999 trials. Average with apparent error of original regression. This approximates the prediction error of the original CER.
$\sigma_{\text {(BCV) }}^{2}=\left(\Sigma_{\mathrm{i}}\left(\sum_{\mathrm{t}} \Delta_{\mathrm{i}, \mathrm{t}}^{2} / \mathbf{N}_{\mathrm{i}}\right) / / \# \mathbf{I} \mid\right.$
$\sigma^{2}{ }_{(\text {(G32") }}=0.368 \sigma_{(\text {app })}+0.632 \sigma_{\text {(BCV) }}^{2}$
$\mathrm{N}_{\mathrm{i}}=$ \# of times the instrument was used for testing
\#I = Total number of instruments


## Planetary Optical Instrument CER

Sensor Cost $($ FY04\$K $)=276.7$ Mass ${ }^{0.426}$ Power ${ }^{0.414}$ DesignLife ${ }^{0.375}$

$$
\mathrm{R}^{2}=0.76 \quad \mathrm{PE}=0.46 \quad \mathrm{~N}=32
$$




## Earth Orbiting Optical Instrument CER

Sensor Cost $($ FY04\$K $)=980$ Mass $^{0.328}$ Power $^{0.357}$ DataRate ${ }^{0.092}$

$$
\mathrm{R}^{2}=0.89 \quad \mathrm{PE}=0.59 \quad \mathrm{~N}=13
$$

Optical Instrument Sensor Cost, Earth Orbiting



## Schedule Estimating Relationship

Nas

## Schedule (months)

$$
\begin{aligned}
& =\boldsymbol{A}_{\text {Misision Type, Instrument Type) }} * \boldsymbol{C o s t t}^{0.107} * \boldsymbol{E} \\
\mathrm{R}^{2}= & 0.66, \sigma_{\text {Predict }}=0.20, \mathrm{~N}=148
\end{aligned}
$$

where Cost is in FY04\$M and $\boldsymbol{E}$ is lognormal, $\boldsymbol{E}=\exp (\varepsilon)$, where $\varepsilon$ is Normal with mean 0 \& standard deviation $\sigma_{\text {Predict }}$
$\boldsymbol{A}($ Mission Type, Instrument Type $)=$


## JCL Simulation

Goal: Determine the Joint Probability of building instrument below Cost Cap and Schedule Cap

## JCL Simulation

2
Step 1: Run the Cost Estimating Relationship, which yields a Cost S-Curve
Step 2: Draw a Random Cost
Step 3: Plug the Random Cost into the Schedule Estimating Relationship
Step 4: Draw a Random Schedule
Step 5: Repeat 2-4 1000x


# JCL Simulation 

Joint Cost \& Schedule Plot


