Data Collection and Analysis Supporting Defendable Cost Estimates

Arlene Minkiewicz, Chief Scientist PRICE Systems, LLC arlene.minkiewicz@pricesystems.com





- Introduction
- Historical Prospective on Cost Estimation
- Data Driven Estimation and its Importance in the current economic client
- Pilot Case Studies
 - Software Data Collection
 - Automotive Data Collection and Analysis
- Lessons Learned

- Cost estimation in the Aerospace and Defense world emerged as a necessary around the time of World War II
- All sorts of mathematical and experiential models have been applied to the problem with varying degrees of success
- More and more organizations are pushing for models tuned to their specific data and experience
- Many of these organizations lack the infrastructure, processes and tools to collect and analyze project data efficiently
- This paper presents two pilot projects (both on-going) to inject data driven estimation into two very different organizations.



• General H.H. "Hap" Arnold at the end of WWII

"During the war, the Army, Army Air Forces, and the Navy have made unprecedented use of scientific and industrial resources. The conclusion is inescapable that we have not yet established the balance necessary to insure the continuation of teamwork among the military, other government agencies, industry, and the universities. Scientific planning must be years in adva of the actual research and development work."

- Leaders took this to heart state of the art weapons systems are necessary to ensure military superiority
- Beginning of period of time during which the Department of Defense (DOD) is plagued with issues surrounding cost!



Historical Perspective

TruePlanning®

by PRICE® Systems



Optimize tomorrow today.™

5

Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com Data Driven Estimation

- Not a new idea just a new focus on what's behind an estimate
- Regardless of estimating methodology a data driven approach lends a sense of credibility and defendability to any estimate
- A 'data-driven' estimate may be.....
 - Based on data from analogous systems
 - Based on a model using analogous systems to determine model inputs
 - Based on a model calibrated to historical data
- A 'data-driven' estimate must be
 - Transparent as to what data was used
 - Clear about how that data was used



Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com Two Case Study Pilots

- Software data collection and analysis to support more defendable estimate
 - Army Program Executive Office for Simulation, Training, and Instrumentation (PEO STRI)
 - No data collection processes in place
 - Pilot involves one large, complex program two contractors one does primarily development, one primarily integration
- Data collection and analysis to support more informed decisions for an automotive manufacturer
 - Low volume auto maker

TruePlanning®

- Collecting tons of cost and technical data at the part level
- Need processes and tools to make sense of that data and mine it into usable form
- Pilot involves about 100 cars of 9 different models

Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com Software Pilot - Motivation

- PEO STRI personnel take their mission of delivering training to the warfighter seriously and intend to deliver best of breed solutions.
- Estimates presented to the Office of the Deputy Assistant Secretary of Defense for Cost and Economics (ODASA-CE) are met with the question "Where's the data that supports this estimate?"
- Unsatisfactory responses inspire large risk margins resulting in less budget available to the program
- Furthermore, without insight into projected costs for specific capabilities, it is difficult to make trade-offs to accomplish reduced costs
- PEO STRI wants to establish a consistent, repeatable process to collect cost and technical data for each capability they develop in their programs

T**rue**Planning®



Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com Software Pilot – Implementation and Process

- Lots of collaboration between PEO STRI personnel, contractors and the PRICE team resulting in.....
 - Integrated Product Team (IPT) developed
 - Data Item Dictionary created to define....
 - What data to collect
 - How often to collect it

uePlanning®



- How to align project specific data to more general categories (across multiple contractors)
- Requirement added to Contract Deliverable Requirements List (CDRL) making data collection a contractor requirement
- Creation of automated code counting tools by wrapping USC's Code Counting tool with Excel
- Software Resource Report (SRR) developed to align with output of code counting tool and automate mapping processes

Software Pilot – Software Resource Report

| | | (Column Will not print) | | | | | | | |
|--|--|---------------------------|---|---------------------------|-------------------------------|--------------------------------|----------------------------------|---|--|
| SECTION 2.5.1 Capability Requirement Name | SECTION 2.5.2 Standardized Capability Name | SECTION 2.5.3 Language | SECTION 2.5.4 % Auto Generated | SECTION 2.5.5 New Code | SECTION 2.5.6 Deleted Code | SECTION 2.5.7 Modified Code | SECTION 2.5.8 Unmodified Code | SECTION 2.5.9 Application Type | SECTION 2.5.10 Estimated % of Total Effort (Build) |
| CR 65774 Reduced System Footprint - SCAMT Multiple Applications on Single Node | Complex Capability | Java | | 1,891 | 3,740 | 628 | 2,492,491 | Assessment/Analysis Functions - Simulation (Constructive) | 12.00% |
| CR 65774 Reduced System Footprint - SCAMT Multiple Applications on Single Node | Complex Capability | XML | | 63 | 36 | 36 | 8,242,067 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67136 Crowd Formation | Moderate Capability | Java | | 978 | - | - | 2,499,444 | Assessment/Analysis Functions - Simulation (Constructive) | 3.00% |
| CR 67136 Crowd Formation | Moderate Capability | XML | | 779 | - | - | 8,208,573 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67136 Crowd Formation | Moderate Capability | XLS File | | 121 | - | - | 815,757 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67136 Crowd Formation | Moderate Capability | CSV file | | 7 | - | - | 276,976 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67154 Dynamic UHRB Aperture Consistent Ordering | Complex Capability | C++ | | 12 | 41 | 34 | 673,807 | Assessment/Analysis Functions - Simulation (Constructive) | 2.00% |
| CR 67154 Dynamic UHRB Aperture Consistent Ordering | Complex Capability | Java | | 238 | 87 | 98 | 2,496,600 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67156 Effects Guidance Matrices | Moderate Capability | Java | | 1,791 | 10 | 15 | 2,497,419 | Assessment/Analysis Functions - Simulation (Constructive) | 35.00% |
| CR 67156 Effects Guidance Matrices | Moderate Capability | XML | | 7 | - | 3 | 8,209,143 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67156 Effects Guidance Matrices | Moderate Capability | XLS File | | 14,066 | - | - | 813,585 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67156 Effects Guidance Matrices | Moderate Capability | CSV file | | 24 | - | - | 276,972 | Assessment/Analysis Functions - Simulation (Constructive) | |
| CR 67164 Raytrace Feature Composition | Complex Capability | C++ | | 982 | 4 | 35 | 676,299 | Assessment/Analysis Functions - Simulation (Constructive) | 13.00% |



Software Pilot – Progress to Date

- Data collection occurs every 10 weeks.
- The first iteration was a thrown away
 - Automation failures in the contractor configuration
 - Uncertainty of some data definitions
 - Additional discussions required by IPT to iron out questions and uncertainties
- Three successful iterations have occurred since then each resulting in slight refinement to the process, DID and/or SRR
- Some practice calibration exercises have led to additional changes to data collection process and the DID
- Current focus is on determination of best long term strategy for storage of this data so it can be accessed and analyzed on a per capability basis



Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com Software Pilot – Lessons Learned

- Entire team needs to understand motivation for data collection
 - Alleviate contractor concern about being 'measured'
 - Facilitate determination of data collection targets
 - Need buy in from the data collectors
- Automation is key to success
 - Creates consistency
 - Creates efficiency
 - Reduces impact on contractors
 - Promotes harmony

"uePlanning"

- Communication and teamwork are paramount
- Flexibility necessary from both the contractors and PEO STRI
 - Not everything is possible or practical
 - Compromises are sometimes necessary



Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com **Automotive Pilot - Motivation**



- Auto manufacturer makes low volume luxury cars and has a vast quantity of cost and technical data for each automobile at the part level.
- This data is collected by various systems and is stored in various files and formats
 - The auto manufacturer is currently unable to predict the cost of an a new or modified vehicle until about two months before the first one rolls out of the plant
- They would like to be able to predict at concept what the cost of the new or modified automobile will be using a question based estimating approach



Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com **Automotive Pilot – Implementation and Process**

- Initial meetings with customer were necessary to understand their mission and identify sources of potential data
- Best potential cost drivers were identified to be the feature codes associated with each individual vehicle
- The sub assembly level was identified as the level for analysis (e.g the seat, dashboard, center console)
- Automation was created to

TruePlanning®

- Align cost and part information at the sub-assembly level
- Align costs with feature codes
- Add visualization capabilities





Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com Data collection and visualization



Optimize tomorrow today.™

TruePlanning[®]

by PRICE® Systems

Presented at the 2013 ICEAA Professional Development & Training Workshop - www.iceaaonline.com **Automotive Pilot – Progress to Date**

- One sub-assembly was identified for the pilot analysis and data for 100 vehicles was collected
 - Automation was applied to extract cost and part information for each vehicle into a single file.
 - Each sub-assembly has a finite number of possible feature codes associated with it
 - Each of these needed to be identified and searched for in the individual files associated with the vehicle Identification Number(VIN). Automation was applied to create mappings between feature codes and costs.
- Data mining techniques were used to identify likely trends
- Regression techniques are currently being applied to create the CERs underlying the question based estimating process
- Further automation and process change being considered to align the disparate data collection systems in the plant.



Data Analysis

TruePlanning[®]

by PRICE® Systems

| <u>F</u> ile <u>E</u> dit | Process 1 | ools <u>V</u> iew | <u>H</u> elp | | | | | | | | | | |
|---------------------------|------------|-------------------|---------------|----------------|---------------|-------------|-----------------|---------------|---------------|----------------|-----------------|----------------|---|
| P 🖄 | - 6 | 🔊 🕼 | A | | 🔊 💆 | 1 | | | | | | | |
| 🛛 🛒 Resul | t Overview | 🗶 🗍 👫 Attri | buteWeights (| Correlation Ma | atrix) 🛛 🗍 🕻 | Correlation | Matrix (Correl | ation Matrix) | 🗶 🚺 Exa | ampleSet (Sele | ect Attributes) | × | |
| Table Vie | w 🔘 Pairw | ise Table 🔘 | Plot View 🔘 | Annotations | | | | | | | | | |
| Attributes | Make | Nameplate | Cost | Seat Materia | I Traffic Msg | . Headlampr | . Front Seatj . | Driver Seat | . Passenger . | Rear Headr | . Rear Cener. | . Psg Slide a. | |
| Make | 1 | 0.733 | 0.669 | 0.424 | 0.511 | 0.569 | 0.431 | 0.219 | 0.219 | 0.378 | 0.270 | 0.666 | P |
| Vameplate | 0.733 | 1 | 0.948 | 0.434 | 0.545 | 0.688 | 0.424 | 0.258 | 0.258 | 0.365 | 0.230 | 0.570 | |
| Cost | 0.669 | 0.948 | 1 | 0.322 | 0.490 | 0.660 | 0.473 | 0.324 | 0.324 | 0.330 | 0.328 | 0.482 | |
| Seat Materia | 0.424 | 0.434 | 0.322 | 1 | 0.375 | 0.277 | -0.163 | -0.040 | -0.040 | 0.108 | -0.055 | 0.225 | Ţ |
| Fraffic Msg C | 0.511 | 0.545 | 0.490 | 0.375 | 1 | 0.120 | 0.174 | -0.195 | -0.195 | 0.206 | 0.036 | 0.301 | |
| -leadlampr (| 0.569 | 0.688 | 0.660 | 0.277 | 0.120 | 1 | 0.100 | 0.350 | 0.350 | 0.224 | 0.431 | 0.408 | |
| Front Seatj A | 0.431 | 0.424 | 0.473 | -0.163 | 0.174 | 0.100 | 1 | 0.544 | 0.544 | 0.185 | 0.314 | 0.239 | |
| Driver Seat A | 0.219 | 0.258 | 0.324 | -0.040 | -0.195 | 0.350 | 0.544 | 1 | 1 | 0.180 | 0.576 | 0.146 | |
| Passenger § | 0.219 | 0.258 | 0.324 | -0.040 | -0.195 | 0.350 | 0.544 | 1 | 1 | 0.180 | 0.576 | 0.146 | |
| Rear Headre | 0.378 | 0.365 | 0.330 | 0.108 | 0.206 | 0.224 | 0.185 | 0.180 | 0.180 | 1 | 0.406 | 0.473 | |
| Rear Cener | 0.270 | 0.230 | 0.328 | -0.055 | 0.036 | 0.431 | 0.314 | 0.576 | 0.576 | 0.406 | 1 | 0.180 | |
| sg Slide an | 0.666 | 0.570 | 0.482 | 0.225 | 0.301 | 0.408 | 0.239 | 0.146 | 0.146 | 0.473 | 0.180 | 1 | |
| Driver Seat T | 0.671 | 0.562 | 0.595 | 0.237 | 0.099 | 0.542 | 0.576 | 0.740 | 0.740 | 0.359 | 0.628 | 0.401 | |
| 'sgr Seat T€ | 0.732 | 0.594 | 0.601 | 0.289 | 0.180 | 0.547 | 0.537 | 0.655 | 0.655 | 0.363 | 0.566 | 0.416 | |
| rogrammal | 0.737 | 0.557 | 0.490 | 0.346 | 0.377 | 0.475 | 0.394 | 0.161 | 0.161 | 0.334 | 0.199 | 0.402 | Ţ |
| Rear Seat B | 0.217 | 0.256 | 0.351 | -0.068 | -0.275 | 0.347 | 0.504 | 0.827 | 0.827 | 0.266 | 0.691 | 0.144 | 1 |
| ingine & Tra | 0.254 | 0.465 | 0.459 | 0.206 | 0.399 | 0.004 | 0.159 | -0.272 | -0.272 | 0.118 | -0.335 | 0.169 | |
| Aulti Functin | 0.519 | 0.529 | 0.520 | 0.304 | 0.810 | 0.097 | 0.229 | -0.095 | -0.095 | 0.149 | 0.086 | 0.346 | |
| Vavigation C | 0.314 | 0.301 | 0.345 | 0.008 | -0.099 | 0.184 | 0.473 | 0.309 | 0.309 | 0.049 | 0.115 | 0.209 | |
| udio Disc S | 0.558 | 0.496 | 0.415 | 0.283 | 0.285 | 0.300 | 0.299 | 0.122 | 0.122 | 0.457 | 0.150 | 0.559 | P |
| Ovd Screen | 0.331 | 0.409 | 0.335 | 0.366 | 0.530 | -0.176 | 0.112 | -0.353 | -0.353 | 0.153 | -0.435 | 0.220 | |
| 3lue Tooth | 0.700 | 0.489 | 0.449 | 0.340 | 0.336 | 0.381 | 0.317 | 0.212 | 0.212 | 0.140 | 0.145 | 0.467 | |
| Sarage Doo | 0.639 | 0.467 | 0.338 | 0.330 | 0.269 | 0.190 | 0.235 | -0.064 | -0.064 | 0.221 | -0.349 | 0.524 | |



- Automation certainly key to success
 - First pass of data mining was done virtually manually.
 Addition of automation pares a weeks worth of work down to seconds.
- Communication
 - Understand the mission of the data collection project
 - Understand the lexicon of the customer
- Analysis of large quantities of data requires visualization and data mining capabilities. Tools such as Rapid-Miner, R and Excel are essential
- For data driven estimating quality of data is more important than quantity – too much data can create noise and misleading results
- Flexibility

T**rue**Planning®



- The need for data driven estimation is driving many organizations in Aerospace and Defense and industry in general to look for new ways to collect and use data
- Data driven estimation is really nothing new we have been using data (real or experiential) to justify our estimates forever. The new requirement is transparency and openness about the data and how it is being used.
- To support this effort organizations may need to establish formal processes for collecting their project data.
- When doing this it is important to understand the importance of
 - Communication
 - Automation
 - Flexibility
 - Buy in across all the stakeholders



Arlene.minkiewicz@pricesystems.com





