

# Enhanced Visualization of Production Systems Concepts and Simulation Data for the Smart/Digital Factory

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## ABSTRACT

Aircraft maintenance, repair, and overhaul (MRO) ensures the continuing airworthiness of aircraft or aircraft parts. Aircraft MRO is extremely costly: the U.S. Air Force spends most of its lifecycle costs to sustain, versus acquire, its aircraft. A key component of Smart Manufacturing for MRO is the use of new software tools for analyzing and exploiting manufacturing knowledge and data more effectively.

This abstract describes two new analytic tools developed for the U.S. Air Force, enhanced with data visualizations, to enable more effective MRO production systems design and operations.

**Selecting appropriate mitigations to production disruptions is challenging because the relationship between control decisions and the resulting performance is delayed and non-linear. The Short-Term Flow Planning Tool helps analysts understand and plan for current and hypothetical changes and disruptions, such as equipment outages, by simulating the effects of various capacity adjustments, induction rates, part scrap rates, and buffer sizes on WIP, part processing rates, and part transfer rates for each process step and day during the planning period. To help analysts interpret this multivariate data, the Tool provides a set of complementary data displays, including heat maps (**

Figure 1), arrays of timelines and time series graphs (Figure 2), and animated bar graphs (Figure 3) which show snapshots of daily production status and performance.

The operations management research literature provides many insights into manufacturing system behavior as well as methods for improving system performance. Unfortunately, many of these results have not been used by practitioners.

**Based on operations management principles, the Diagnostic Tree (D-tree) provides a systematic way of reviewing possible ways of improving a production system. A D-tree is composed of nodes which each describe a performance improvement objective. Many of the nodes are linked to sub-nodes that describe more concrete sub-objectives for achieving the objective and describe how the objective's metric varies with the values of its sub-objective metrics. To help analysts understand these multivariate relationships, a sensitivity analysis function at selected nodes shows how the objective function varies with a user-selected variable, fixing the other variables at user-specified values. The dialog in**

**Figure 4 prompts the user for values of the sub-objectives, and the graph in** shows how the objective variable varies with the range variable -- in this case, sequential process batch size.

## ACKNOWLEDGEMENT

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|    | A                      | B   | C   | D   | E   | F   | G   | H   | I   | J   | K    | L    | M    | N    | O    | P    |
|----|------------------------|-----|-----|-----|-----|-----|-----|-----|-----|-----|------|------|------|------|------|------|
| 1  | <b>Parts Processed</b> |     |     |     |     |     |     |     |     |     |      |      |      |      |      |      |
| 2  |                        | 1   | 2   | 3   | 4   | 5   | 6   | 7   | 8   | 9   | 10   | 11   | 12   | 13   | 14   |      |
| 3  |                        | 1/1 | 1/2 | 1/3 | 1/4 | 1/5 | 1/6 | 1/7 | 1/8 | 1/9 | 1/10 | 1/11 | 1/12 | 1/13 | 1/14 | Sum  |
| 4  | Step 1                 | 12  | 12  | 12  | 12  | 12  | 4   | 0   | 12  | 12  | 12   | 12   | 12   | 12   | 12   | 148  |
| 5  | Step 2                 | 12  | 12  | 12  | 12  | 12  | 0   | 0   | 12  | 12  | 12   | 12   | 12   | 12   | 12   | 144  |
| 6  | Step 3                 | 12  | 12  | 12  | 12  | 8   | 0   | 0   | 12  | 12  | 12   | 12   | 12   | 12   | 12   | 140  |
| 7  | Step 4                 | 12  | 12  | 12  | 0   | 0   | 0   | 16  | 16  | 12  | 12   | 12   | 12   | 12   | 12   | 140  |
| 8  | Step 5                 | 12  | 12  | 12  | 12  | 0   | 0   | 0   | 12  | 12  | 12   | 12   | 12   | 12   | 12   | 132  |
| 9  | Step 6                 | 12  | 12  | 12  | 12  | 12  | 0   | 0   | 0   | 12  | 12   | 12   | 12   | 12   | 12   | 132  |
| 10 | Step 7                 | 12  | 12  | 12  | 12  | 12  | 12  | 0   | 0   | 0   | 12   | 12   | 12   | 12   | 12   | 132  |
| 11 | Step 8                 | 12  | 12  | 12  | 12  | 12  | 12  | 0   | 0   | 0   | 12   | 12   | 12   | 12   | 12   | 132  |
| 12 | Step 9                 | 12  | 12  | 12  | 12  | 12  | 12  | 12  | 0   | 0   | 0    | 12   | 12   | 12   | 12   | 132  |
| 13 | Step 10                | 12  | 12  | 12  | 12  | 12  | 12  | 12  | 12  | 0   | 0    | 0    | 12   | 12   | 12   | 132  |
| 14 | Sum                    | 120 | 120 | 120 | 108 | 92  | 52  | 52  | 88  | 84  | 84   | 96   | 108  | 120  | 120  | 1364 |

Figure 1 – Short-Term WIP Flow Tool Heat Map Shows Parts Processed for Each Step and Day

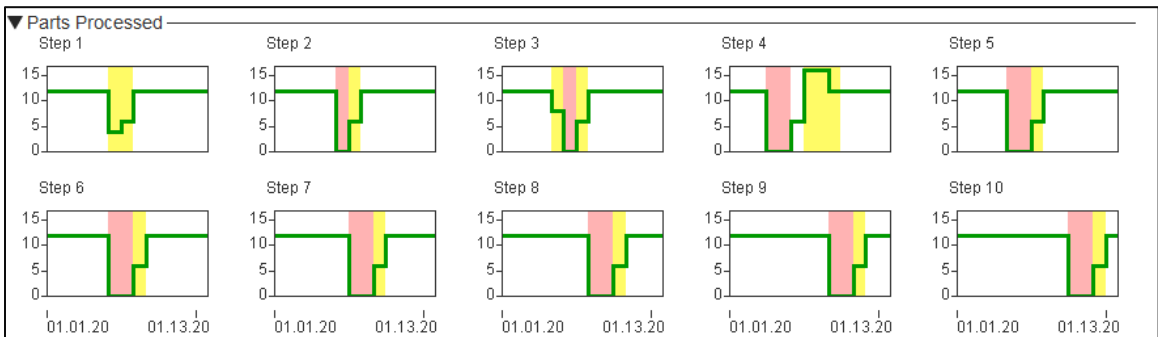


Figure 2 – WIP Flow Tool Graphs Shows Production Variable Values by Step vs. Time.

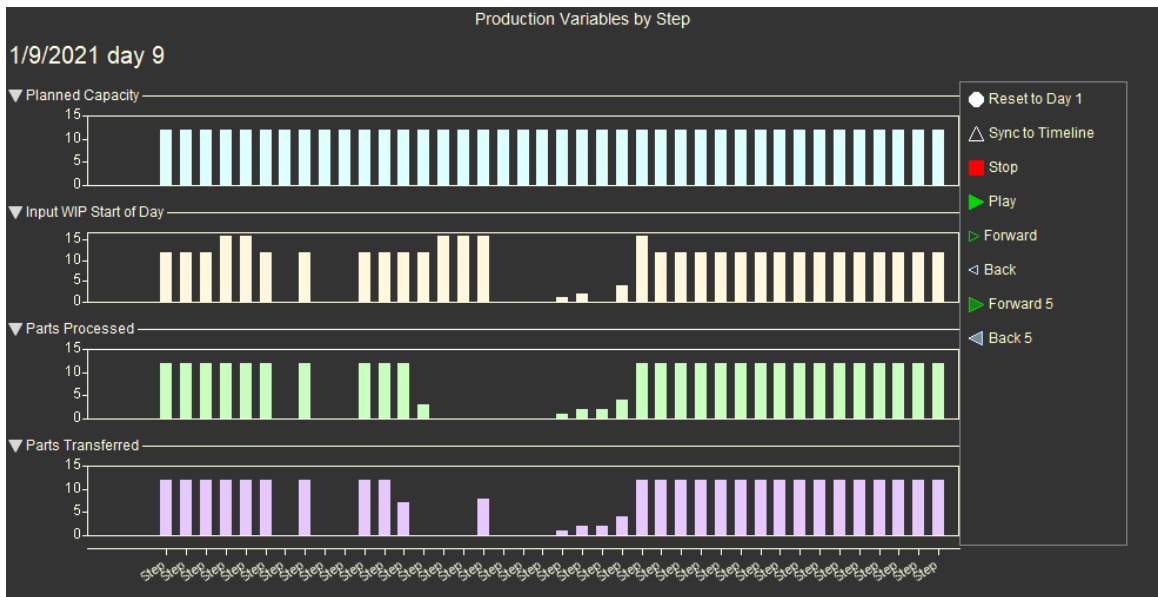


Figure 3 – Animated Bar Graphs Show Daily Status and Performance by Step vs. Day

Diagnostic Tree

Node Browser
Estimator \*
Node Listing
All Comments
Help

### Adjust Batch Size

In a sequential batch process, the cycle time of station i equals:

$$CT_i = CT_q + s + WIBT + t$$

where:

- $CT_q$  = station queue time =  $(c_a^2 + c_e^2)/2 * u/(1-u) * t_e$
- WIBT = Wait in Batch Time =  $(k_p-1) * t/2 + (k_t-1) * t/2$
- $u$  = station utilization =  $r_a * t_e / k_p$
- $t_e$  = effective process time for a batch =  $k_p * t + s$

|         |  |   |
|---------|--|---|
| 5 to 20 |  | $k_p$ size of the sequential processing batch ( $x \geq 1$ )  |
| 1       |  | $s$ setup time ( $x \geq 0$ )   |
| 1       |  | $t$ station process time for a single part ( $x \geq 0.0001$ )  |
| 0.8     |  | $r_a$ average rate of parts arrivals at the station ( $x \geq 0$ )  |
| 1       |  | $c_e$ coefficient of variation (CV) of the effective process time of a batch, including both process time and setup time ( $x \geq 0$ ) |
| 1       |  | $c_a$ coefficient of variation (CV) of the times between part arrivals at the station ( $x \geq 0$ )                                    |
| 1       |  | $k_t$ size of the transfer batch ( $x \geq 1$ )   |

|  |   |
|--|---|
|  | $CT_i$ average time a part spends at each station, including processing, queueing, and waiting time |
|--|---|

Figure 4 –D-Tree Estimator Inputs Specify a Range for Batch Size ( $k_p$ )

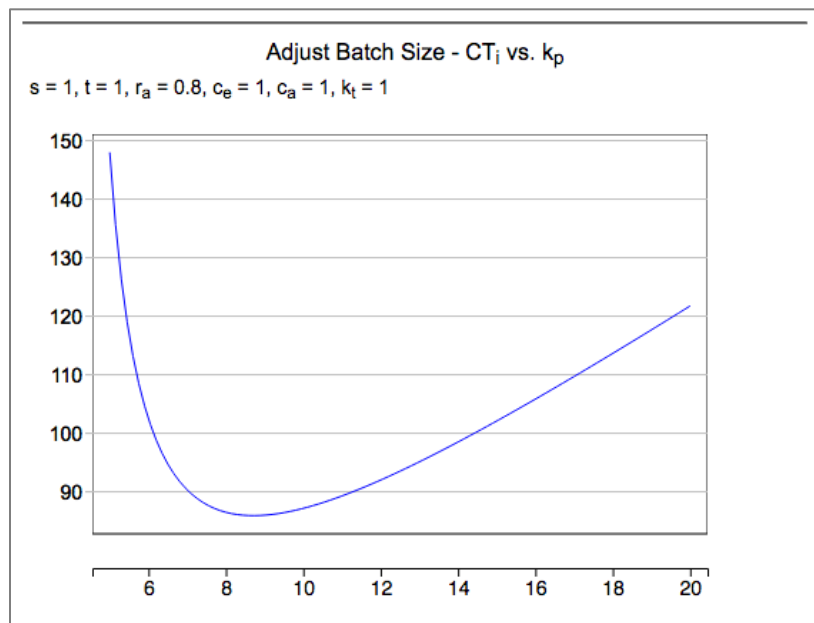


Figure 5 – D-Tree XY Graph Shows How Station Cycle Time Varies with Batch Size at User-Specified Values of other Variables