## DELIVERY CONFIDENCE MODEL

Forecasting Projected Throughput in a Complex Assembly Manufacturing System using Neural Networks and Capacity Simulation

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**Operations Analysis - Industrial Engineering** 



### **LOCKHEED MARTIN OVERVIEW**

- One of the world's leading aerospace, defense, security, and technology companies.
- Over 100,000 Employees worldwide.
- Developed and produced iconic military aircraft such as the U-2, SR-71, F-16, C-130, and F-117 stealth fighter.
- Current largest contract is the F-35 Joint Strike Fighter.
  - 5<sup>th</sup> generation, multi-role stealth fighter.
  - CTOL, STOVL, and CV.

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### **MANUFACTURING SYSTEM STUDIED**

#### **System Highlights:**

- Global aeronautics defense production program.
- Currently a low throughput system with a growing customer base and demand for product.
- Thousands of parts and install operations.
- Complex installations with demanding specifications.
- Large scale customer visibility.

## PROJECT BACKGROUND OVERVIEW



### **CASE FOR MODEL CREATION**

#### **Original Question:**

• As of time now, what is the probability that I will hit my annual production throughput goal for the next 3 years?

#### Old answer to that question:

- Take the current production schedule and forecast a performance regression to a recovery point and build a detailed schedule using Microsoft Excel.
  - Manual creation, siloed execution between production areas.
  - Build 1 area's schedule out at a time.
  - No variation in performance is included.

#### **Results leave more questions with unknown answers:**

- You didn't study variation, so does this mean I will make my goal 100% of the time?
- What variables are driving any risks to reaching that goal?
- Which variables have the greatest influence on the output?
- Is the learning curve applied actually feasible?
- Etc.



### WHAT REALLY NEEDS TO BE STUDIED?

#### **Problem Statement:**

• As of time now, what is the probability that I will hit my annual production throughput goal for the next 3 years?

#### **Understanding:**

- Problem is way more complicated than saying this task takes X days and in the future I believe I can perform at Y.
  - Will parts be available, is manpower trained and available, what is my defect rate, etc.?

#### What do I really need in order to solve answer:

- Ability to understand the thousands of complex intervariable relationships and how that effects the production span for a given area.
- Need a integrated schedule tool that understands capacity relationships.
- Needs to study variation in performance.
- Trusted forecasts for all known inputs in the production system that can effect span and the associated change risks.
- Ability to identify risk and tests mitigation strategies and goals prior to implementation.

# STRATEGIES USED TO CREATE MODEL

INTEGRATING NEURAL NETWORKS WITH CAPACITY SIMULATION



### **QUICK GLOSSARY OF TERMS**

#### <u>Terms</u>

- <u>Neural Network</u> An artificial intelligence algorithm that is created based on historical relationships between input variables and a known output, in order to predict a future output with known input variables and an unknown relationship between those variables.
- <u>Markov Chain</u> Is a stochastic model describing a sequence of possible events in which the probability of each event depends only on the state attained in the previous event.
- <u>Deviation Distribution</u> A distribution based on the deviation ratio of the actual value from the expected value.



### **VARIABLES STUDIED IN NEURAL NETWORK**

#### Variables in Delivery Confidence Model

- Material Shortage Forecasts
- Product Variant Mix
- Production Performance
- Scrap, Rework, and Repair
- Out of Station Work quantity and how far it travels downstream
- Staffing Headcount Levels
- Knowledge, Skills, and Abilities of Employees
- In-plant fabrication supplier performance metrics
- Takt Rate
- Span of Individual Areas
- Production Learning Curve Rates
- Continuous Improvement Initiatives
- Seasonality
- Seasonality of Work Force Attendance and Performance Trends



### HOW IS THIS ACCOMPLISHED?





### **SIMIO LOGIC AND USES**

#### Data Driven Model (Data Tables)

- Entity Arrival table loaded with Span Projections using Markov Chain Deviation Distribution from the expected value
- Task Sequences enable more granular modeling and tracking
- Custom work schedules, including planned and unplanned maintenance and downtime by resource

#### **Entity Flow and Logic**

- Routing Groups used to regulate variable entity flow between component area rate stations
- Add-on Processes enable custom scenario routing logic based on inputs and model conditions

#### Model Validation and Results Analysis

- Resource and Entity Gantt Charts utilized for quick model troubleshooting and validation
- Write out steps used to capture key metrics and load into custom output visuals:
  - Milestone start and complete dates for each task and resource
  - Resource utilization analysis
- Experiments and Operational Planning (Risk Analysis) used to build confidence intervals for given scenarios





### **SCENARIO ANALYSES STUDIED**

	<u>Increased</u> <u>Throughput</u>	<u>Capacity Planning</u>	<u>Risk</u> Identification	<u>Risk Mitigation</u>
SPAN/SCHEDULE	Annual Year over year Increase in Production Demand	Gain Insight in regards to adequate facility capacity vs. WIP Projection?	Need to quickly identify and quantify varying risk impacts outside isolated area.	Need for a tool to understand potential programmatic changes for improvement.
VARIATION	Does span performance variation limit our ability to reach our increasing demand profile?	Does variation or maintenance downtime cause a bottleneck?	Which variables cause the most disruptive and frequent variation to production performance?	What variables should I target with improvement initiatives to reduce variation?
BUSINESS CASE	Missing Demand Profile, means loss in Company Value	ROI of fixing Capacity Bottleneck	More enhanced analyses on what is truly driving production performance risk vs. gut feel.	In depth analytics to accurately reflect risk mitigation strategies and success rates.



## TABLEAU DEMO VIEWER



 $\leftarrow \ {\tt Undo} \ \rightarrow \ {\tt Redo} \ \ \overleftarrow{\leftarrow} \ \ {\tt Revert} \ \ \overleftarrow{{\tt G}} \ \ {\tt Refresh} \ \ \overleftarrow{{\tt G}} \ \ {\tt Pause}$ 

### **PROGRAM LEVEL VIEW**



ANNUAL THROUGHPUT PROBABILITY

92%

#### Forecasted Output:

Category	Area # 1	Area # 2	Area # 3	Area # 4	Area # 5	Area # 6	Area # 7	Area # 8	Area # 9	Area # 10
6 Month Build Span Avg	30	40	50	60	50	40	30	20	10	10
6 Month -12 Month Build Span Avg	28	37	52	61	45	37	33	18	9	8
12 month-2 Years	25	35	47	57	40	35	30	18	9	6
2019 Annual Quantity	Α	В	С	D	Е	F	G	Н	I	J

#### Variables to Toggle (Default Forecast Shown):

Category	Area # 1	Area # 2	Area # 3	Area # 4	Area # 5	Area # 6	Area # 7	Area # 8	Area # 9	Area # 10
Variable # 1	15% 🗨	8% 🗖	18%	23%	20% 🗖	25% 🗖	30% 🗖	12%	15% 🗖	70% 🗖
Variable #2	10	8	2	6	2 🗖	15 🗖	38 🗖	12	3 🔽	1
Variable #3	10	8	2	6	2	15	38	12	3 🗖	1
Variable #4	310	180	137	625	190	379	627	168	98	160
Variable #5	150 🔽	70 🔽	220 🗨	350 🔽	60 🖵	250	460 🗖	N/A	20	N/A
Variable #6	48%	39%	37% 🖵	15% 🗖	35%	17%	14%	29%	30%	51%
Variable #7	100%	105% 🗖	90% 🔽	100% 🖵	110% 🖵	95% 🗨	95%	100%	95% 🗨	90%
Variable #8	58 💌	18 🔽	24 🔽	25 💌	66 🗸	21	30 🗸		10 🔽	26 🔽



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### **Component Area:** AREA # 1

#### **Top Correlated Variables to Output**

Top Correlated Variables to Span	Value	Span Impact
Variable # 37	.758	.05 days/event
Variable # 33	.645	.12 days/event
Variable #25	.412	.08 days/event
Variable # 2	.432	.5 days/event
Variable #16	.097	Not Correlated
Etc.	Х	N/A

#### **Potential Impact at Perfection:**

Variable #37 = 5 Days savings Variable #33 = 3 Days savings Variable #25 = 2 Days savings Variable #2 = 1 Days savings

#### Variables Correlation Heat Map



### SIMID GANTT CHART OF DETAILED SCHEDULE

\*DATA IS NOTIONAL

ProcessA09	Process E03     Process E03     Process E01     Process E01	
ProcessA03	Process 608         Process C07         Process         Process F03	
ProcessA08	Process B14         Process C08         Process F07	
ProcessA02	Process B20 Proces	
ProcessA05	ProcessB02 ProcessB02 ProcessC06 Pro ProcessB02 ProcessF10	
ProcessA12	ProcessB15 ProcessB15 ProcessC09 ProcessE ProcessE ProcessE11	
ProcessA10	ProcessB24 ProcessB24 ProcessB24 ProcessB203 ProcessB203 ProcessB203 ProcessB203 ProcessB203 ProcessB204 ProcesB204 ProcesB204 ProcesB204 ProcesB204 ProcesB204 Pr	
ProcessA02	Process B21 Process B21 Process B21 Process CD6 Process CD6 Process CD6 Process FD2 Proces	
ProcessA10	ProcessB25         ProcesS25         ProcesS25         ProceS25<	
ProcessA04	ProcessB01 ProcesB01 ProcesB01 ProcessB01 ProcesB01 ProcesB01 ProcesB01 Proce	
ProcessA11	ProcessB06         ProcessC07         ProcessE00         Process	
ProcessA06	Process B09 Proces	
ProcessA01	ProcessB10 ProcesB10 ProcesB10 ProcesB10 ProcesB10 ProceB10	
Proc	Process B22 Process B22	
ProcessA11	A11 ProcessB11 ProcesB11 ProceB11 ProceB1	
Р	ProcessA07         ProcessB26         ProcessB26         ProcessE0	
	ProcessA07         ProcessB28         ProcesB28         ProcesB28         ProcesB28	
	ProcessA09         ProcessB30         ProcessB30         ProcessF04	
	ProcessA12         ProcessB23         ProcessB23         ProcessC12         ProcessF12         ProcessF12	
	ProcessA13 ProcessB13 ProcesB13 Pr	
	ProcessA08         ProcessB07         ProcessB07         ProcessC11         P	



### **BOX-WHISKER VIEW OF DELIVERY TIMELINE FROM SIMIO**

Projected Deliveries Timeframe



Time Product Delivered





## **ROBOT CAPACITY UTILIZATION FORECAST**







### PARETO OF VARIABLE IMPACT TO FINAL SPAN



Neural Network will auto-rank impact variables to Span Output.



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# **QUESTIONS?**



