

Development of a Part Criticality Index in Inventory Management

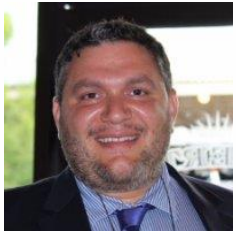
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INTRODUCTION



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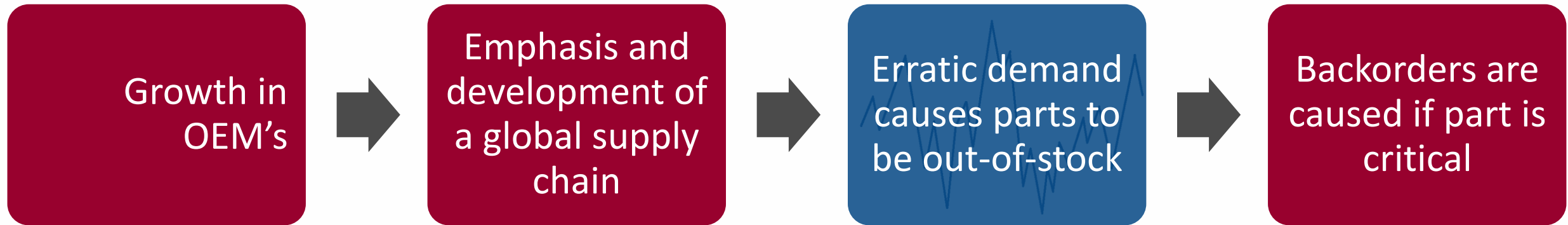
Liudas Panavas

- University of South Carolina McNair aerospace center
- Sophomore undergraduate
- Mechanical engineering

- Talk about what we are going to talk about – outline of presentation



INTRODUCTION



Solution

Optimize restocking process using the concept of part criticality



OBJECTIVES

Determine

- the **parameters** that are the most important in **determining part criticality**.
- the most critical part in order to monitor its **availability in the inventory**.

Ensure

- that inventory is **dynamically evolving**
- a high level of **customer satisfaction**
- the solution is **cost effective**.



ECONOMIC IMPACT

- Bullwhip Effect
 - Backlogs
 - Costs

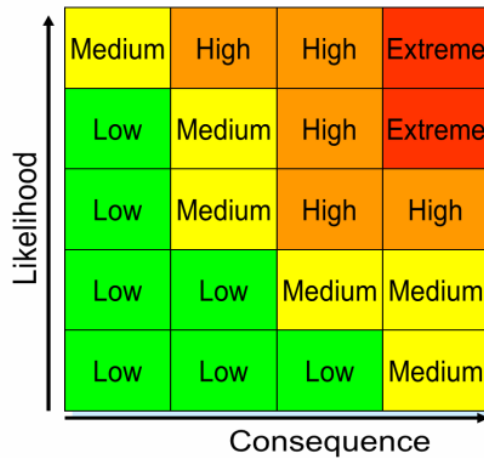
- Cooperation between supplier and manufacturer
- Customer satisfaction



BACKGROUND/LITERATURE

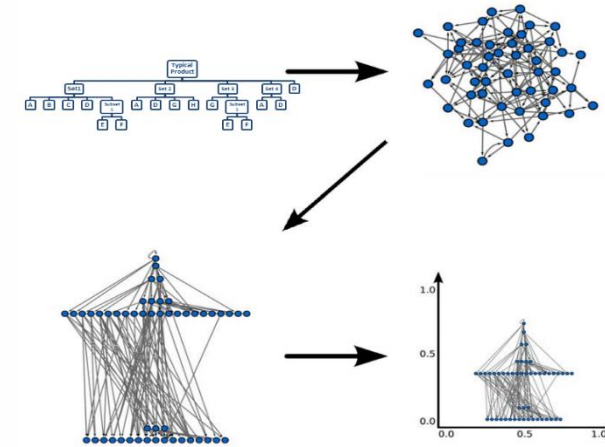
Spare Part Criticality

- Classification of criticalness of parts using **likelihood of failure and consequence of failure**



Hierarchy in complex relationship

- Mones et al. uses nodes and edges to describe **fundamental features of complex systems**

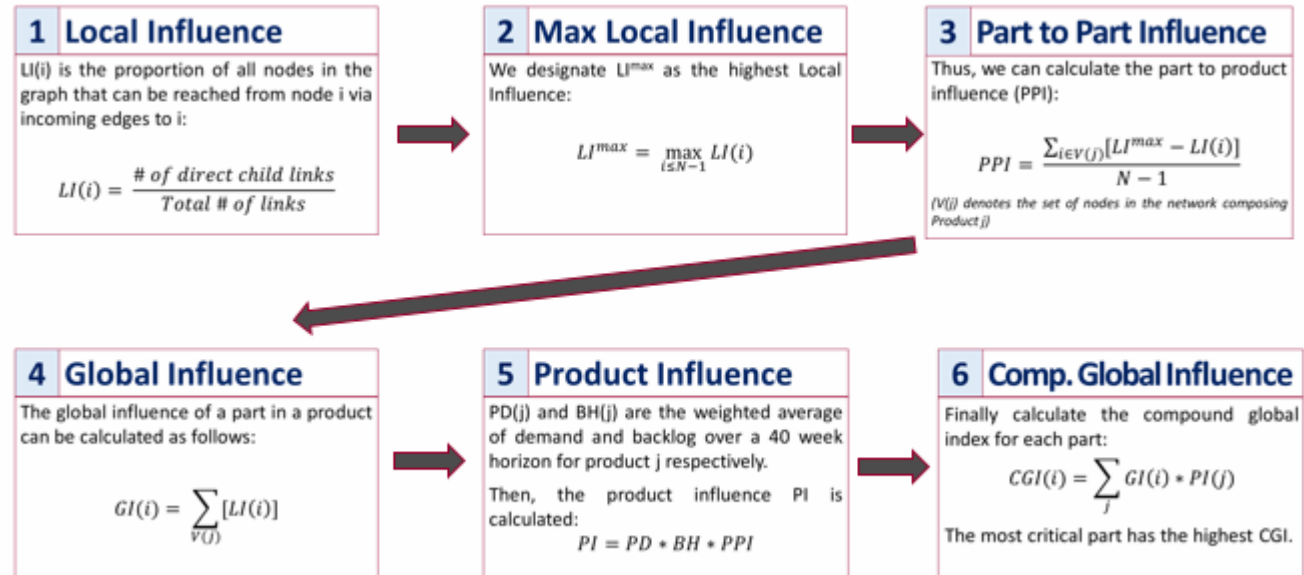
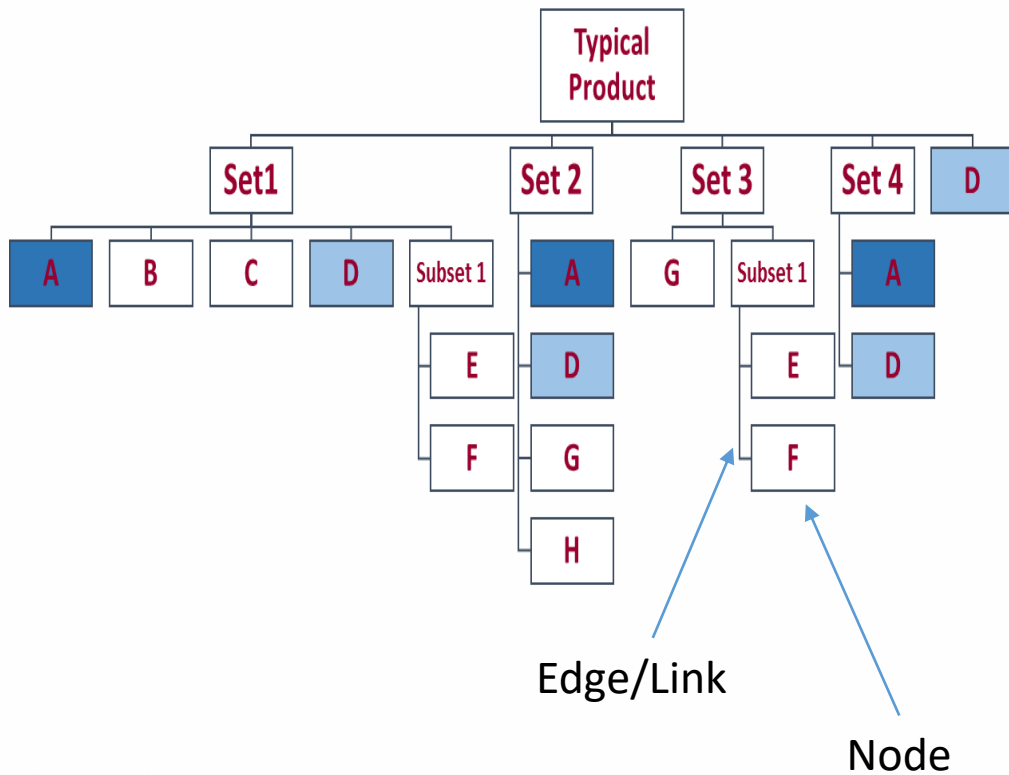


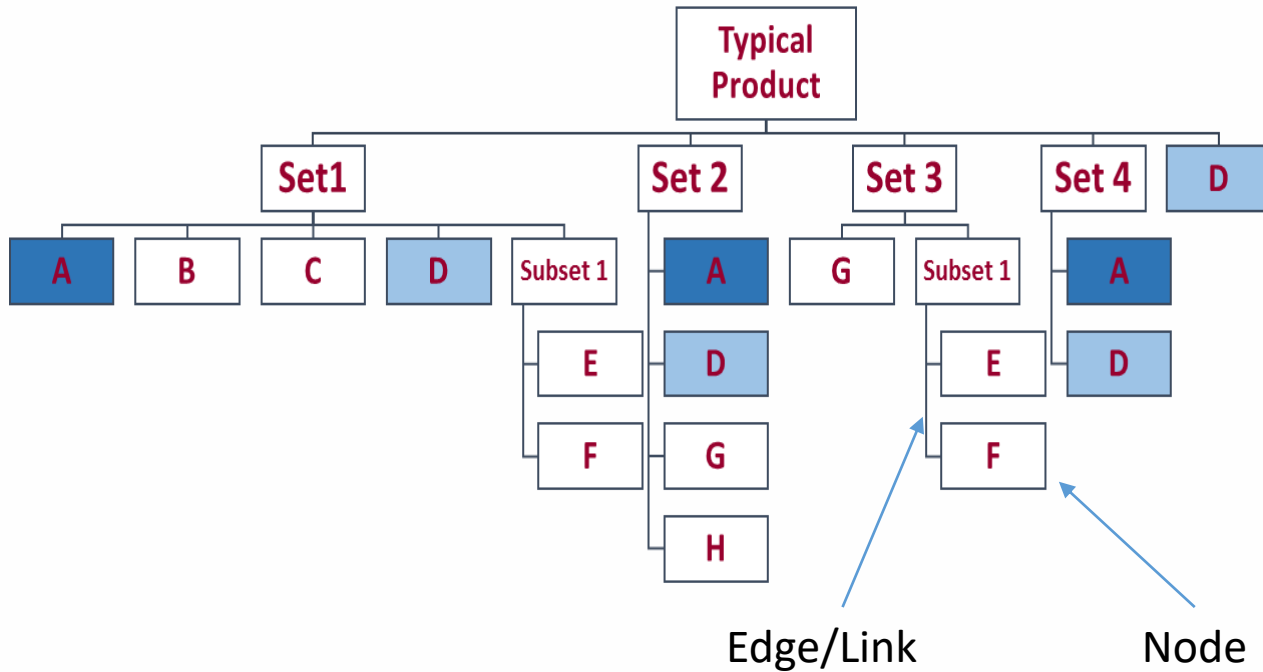
Gap

No system in place to identify criticalness of parts in the complex system of a product



METHODOLOGY





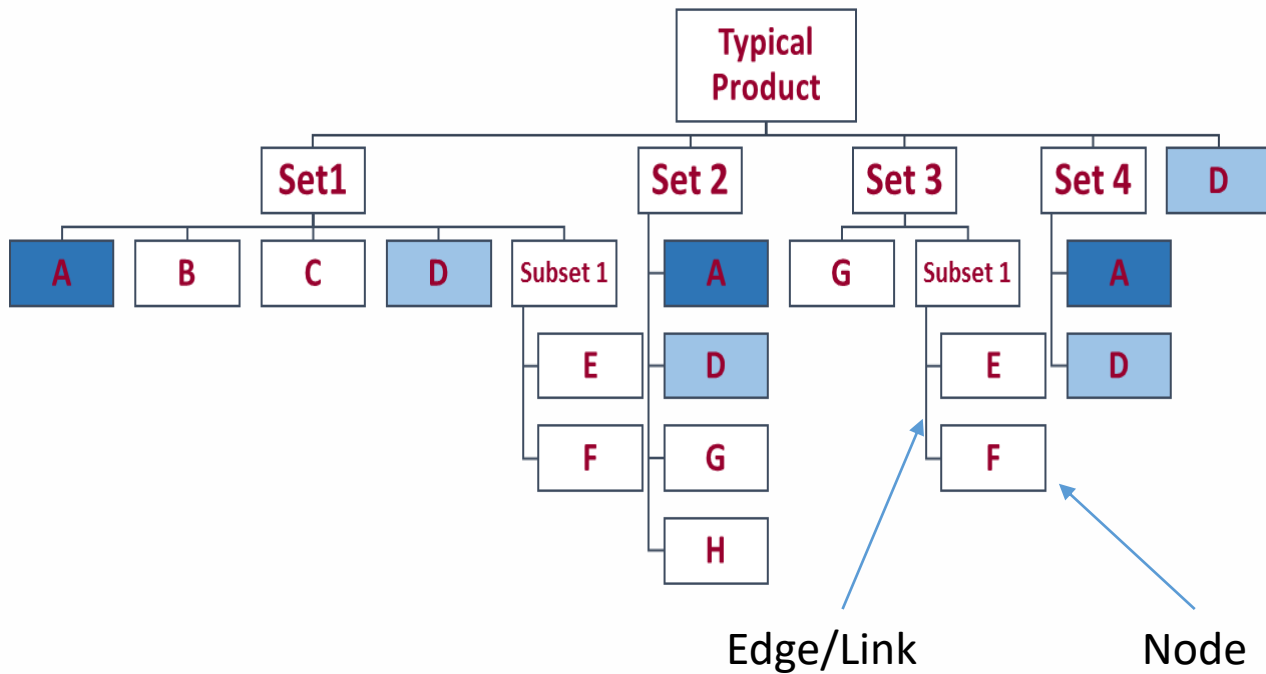
1 Local Influence

$LI(i)$ is the proportion of all nodes in the graph that can be reached from node i via incoming edges to i :

$$LI(i) = \frac{\# \text{ of direct child links}}{\text{Total \# of links}}$$

- # of direct child links is 1 for all parts in our product
- Calculates relative importance of each individual part





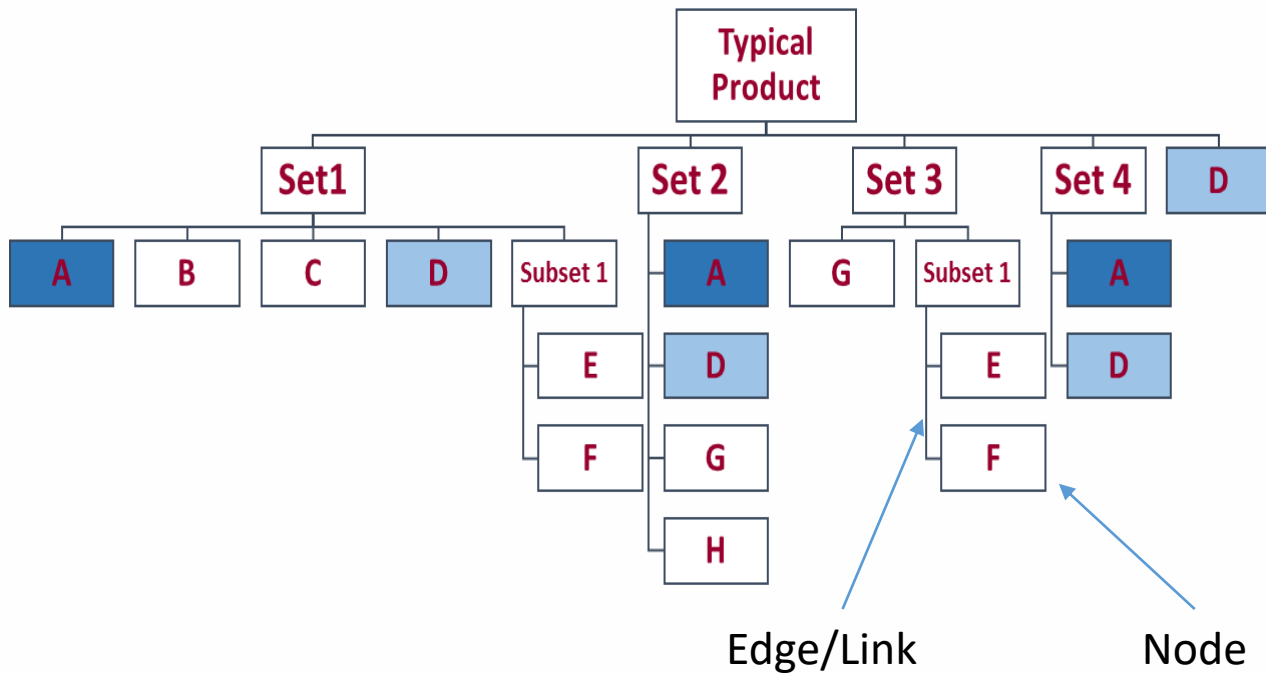
2 Max Local Influence

We designate LI^{\max} as the highest Local Influence:

$$LI^{\max} = \max_{i \leq N-1} LI(i)$$

- Calculates which part has the largest local influence
- Will be used to normalize results when comparing different products





3 Part to Part Influence

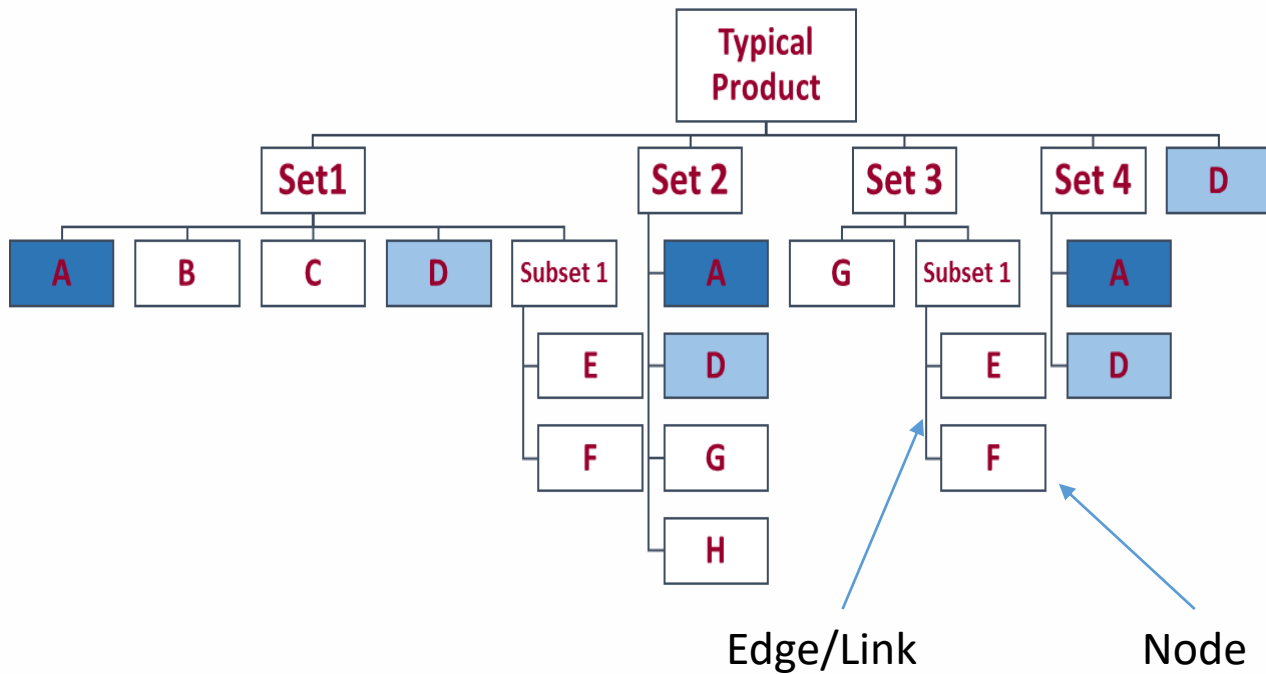
Thus, we can calculate the part to product influence (PPI):

$$PPI = \frac{\sum_{i \in V(j)} [LI^{max} - LI(i)]}{N - 1}$$

(V(j) denotes the set of nodes in the network composing Product j)

- Allows for the comparison of the same part in different products





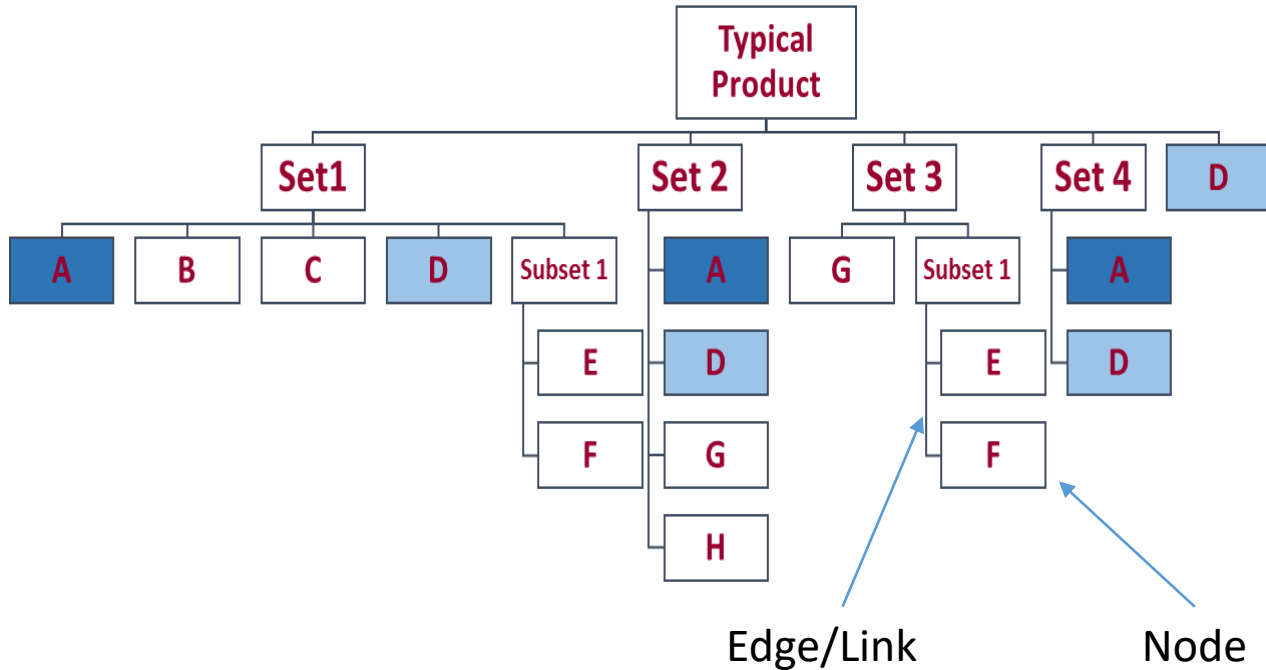
4 Global Influence

The global influence of a part in a product can be calculated as follows:

$$GI(i) = \sum_{v(j)} [LI(i)]$$

- Calculates the cumulative importance of each part in a product





Week	Demand	Supply	Inventory	Backlog
1	10	9	0	1
2	9	10	0	0
3	10	9	0	1
4	10	12	1	0
5	11	10	0	0
6	9	8	0	1
7	10	12	1	0
8	9	12	4	0

5 Product Influence

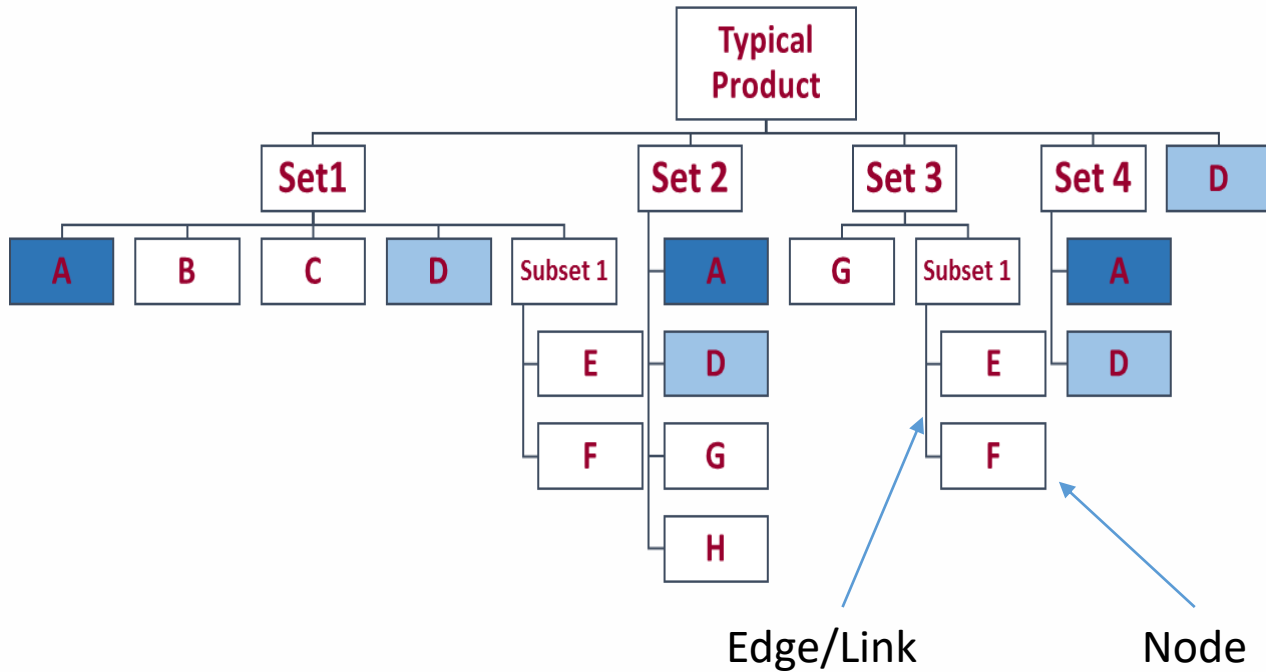
PD(j) and BH(j) are the weighted average of demand and backlog over a 40 week horizon for product j respectively.

Then, the product influence PI is calculated:

$$PI = PD * BH * PPI$$

- Demand and supply constant stochastic values
- Determines which products are most critical by examining product demand, backlog, and part to part influence





6 Comp. Global Influence

Finally calculate the compound global index for each part:

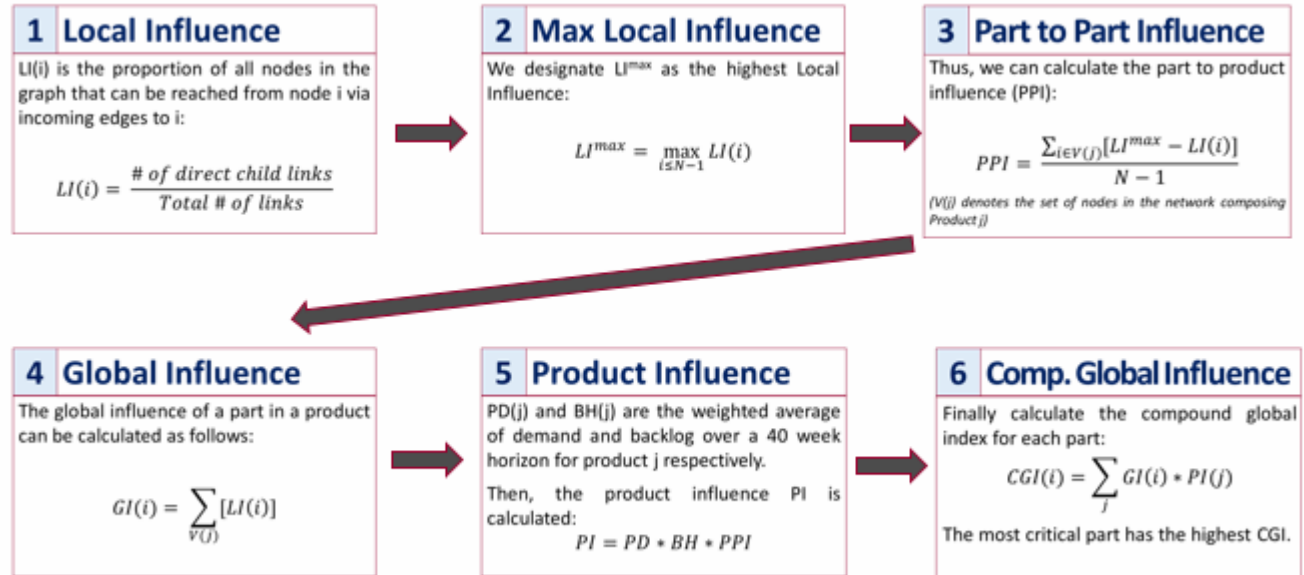
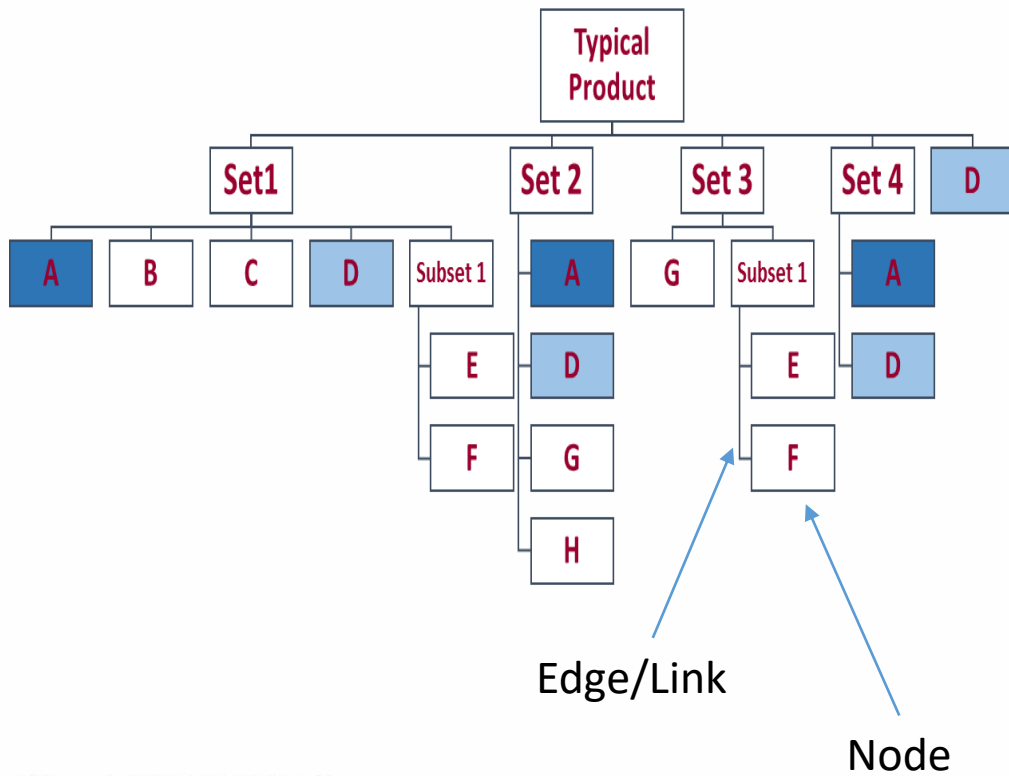
$$CGI(i) = \sum_j GI(i) * PI(j)$$

The most critical part has the highest CGI.

- Connects part criticality to product criticality
- Identifies most important part across all products



METHODOLOGY



RESULTS

- Short run - The highest Compound Global influence **varies depending on run**

Part	Run 1	Run 2	Run 3	Run 4	Run 5	Long Run
Set 1	0.0037707 5	0.0155130 6	0.0027851 3	0.0126266 4	0.0097451 7	0.00762882
Set 2	0.0147669 1	0.0087029 2	0.0063884 3	0.0101218 3	0.0136005 1	0.01741808
Set 6	0.0147669	0.0004902	0.0036951 4	0	0.0118145 8	0.00638000
Set 7	0.0147669	0.0004902	0.0036951 4	0	0.0118145 8	0.00638000
Subset 1	0.0037707	0.0155130	0.0054784 2	0.0147391 8	0.0106516 7	0.01095505
A	0.0336402	0.0340810	0.0217905 7	0.0330991 6	0.0372055 5	0.04400146
B	0.0037707	0.0237257	0.0027851 3	0.0206359 3	0.0106245 8	0.01534066
D	0.0223084	0.0479417	0.0146519 8	0.0454969 5	0.0348767 8	0.04371381
E	0.0037707	0.0155130	0.0054784 2	0.0147391 8	0.0106516 7	0.01095505
F	0.0076533	0.0314087	0.0076360 5	0.0253295 6	0.0195767 9	0.01576854
G	0.0147669	0.0087029	0.0090817 2	0.0122343 8	0.0145070 2	0.02074432
H	0.0148787	0.0090855	0.0084542 1	0.0101981 2	0.0136869 6	0.01792899
J	0.0147669	0.0087029	0.0036951 4	0.0080092 9	0.0126940 0	0.01409185
L	0.0148787	0.0008728	0.0057609 2	7.63E-05	0.0119010 4	0.00689090
M	0.0334164	0.0168612	0.0095479 1	0.0105903	0.0325542	0.01757349
N	0.0443007	0.0014706	0.0110854 3	0	0.0354437 6	0.01914000 8
O	0.0223084	0.0315163 6	0.0038788 3	0.0210281 9	0.0294919 1	0.01498517



RESULTS

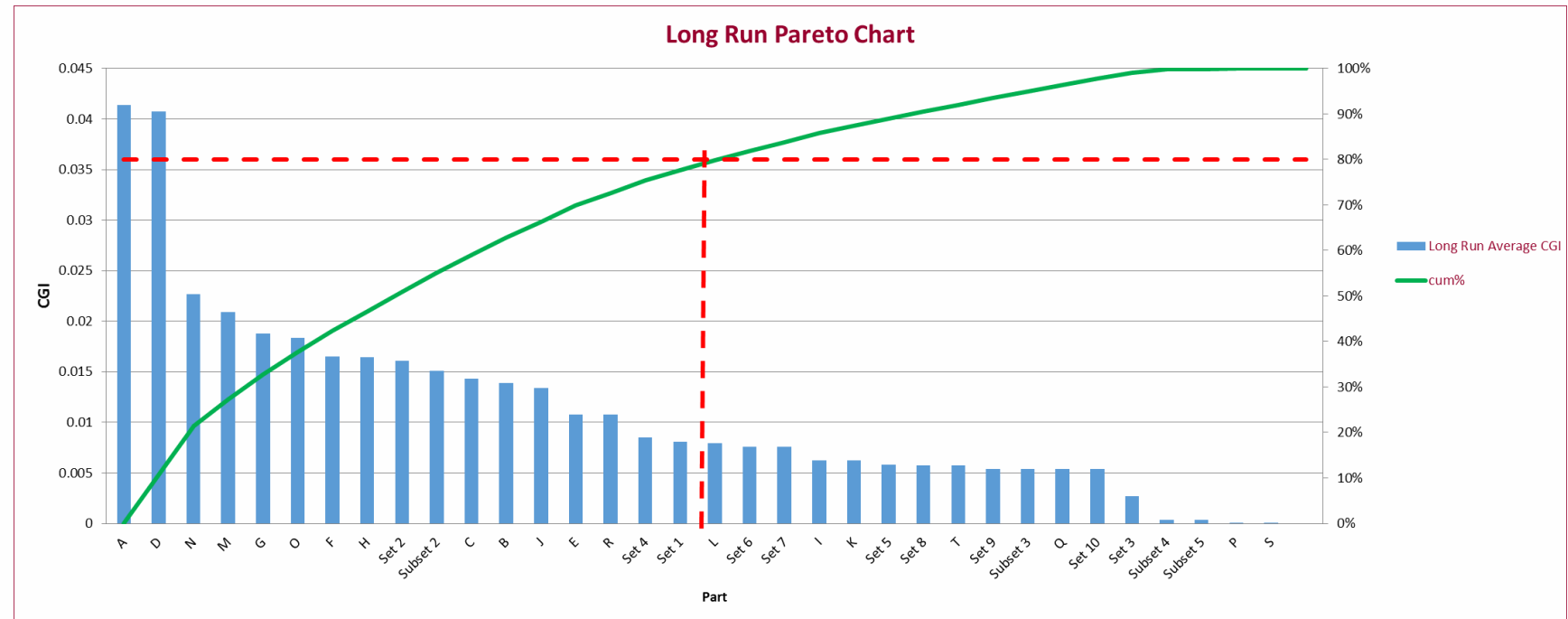
- Short run - The highest Compound Global influence **varies depending on run**
 - Demand and supply are random variables
- No definitive results for most important part

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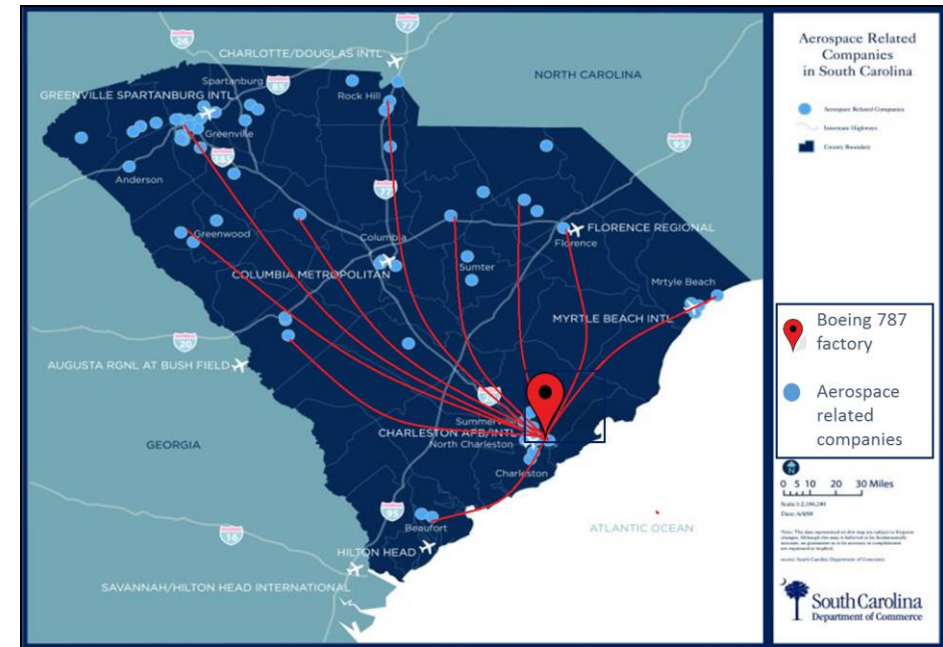
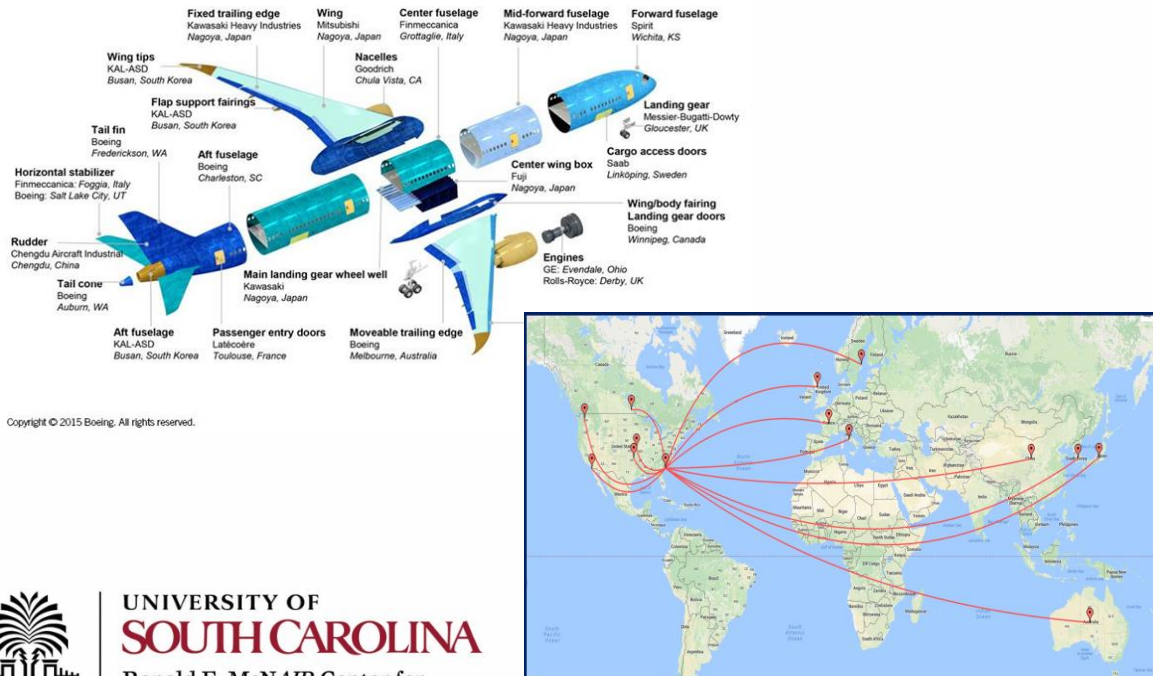
RESULTS

- Long run – **One part** consistently has the **highest criticality**
 - Random variables associated with supply and demand are normalized
- Pareto chart helps display cumulative importance of parts
 - Disobeys the 80/20 rule



INDUSTRY SIGNIFICANCE

- Companies such as Boeing have to ship parts from **all over the world** to create a product
- Missing part can cause **long delays** in the assembly process
- Having a local manufacturer would **reduce time it takes to receive parts**
- **Symbiotic relationship** helping OEM and local manufacturers



BIG PICTURE

Name	Inhibit Type	Address	Address 2	City	State/County/Postal Code	Phone	Fax	Date	Country	Link	Email
Sample Location 1		4234 Sample Way	Sub 2015	Tacoma	WA 98421	252-252-2524			United States	http://www.sample.com	info@sample.com
Sample Location 2		4234 Fake Address Lane		Opalom	MO 62268	252-252-2525			United States	http://www.sample.com	info@sample.com
Sample Location 3		458 W Address Road		Arlington Heights	IL 60005	252-252-2526			United States	http://www.sample.com	info@sample.com
Sample Location 4		789 East Street Way		Austin	TX 78768	252-252-2527			United States	http://www.sample.com	info@sample.com
Sample Location 5		1234 SE MetaLocator Way		Summerfield	FL 34464	252-252-2528			United States	http://www.sample.com	info@sample.com
Sample Location 6		8901 Tester Lane		New Orleans	LA 70130	252-252-2529			United States	http://www.sample.com	info@sample.com

Local part manufacturer database



Algorithm finds critical parts



Relationship is built with manufacturer that can supply critical parts



Critical part missing on assembly line



Order placed for part



Part delivered quickly and assembly can continue



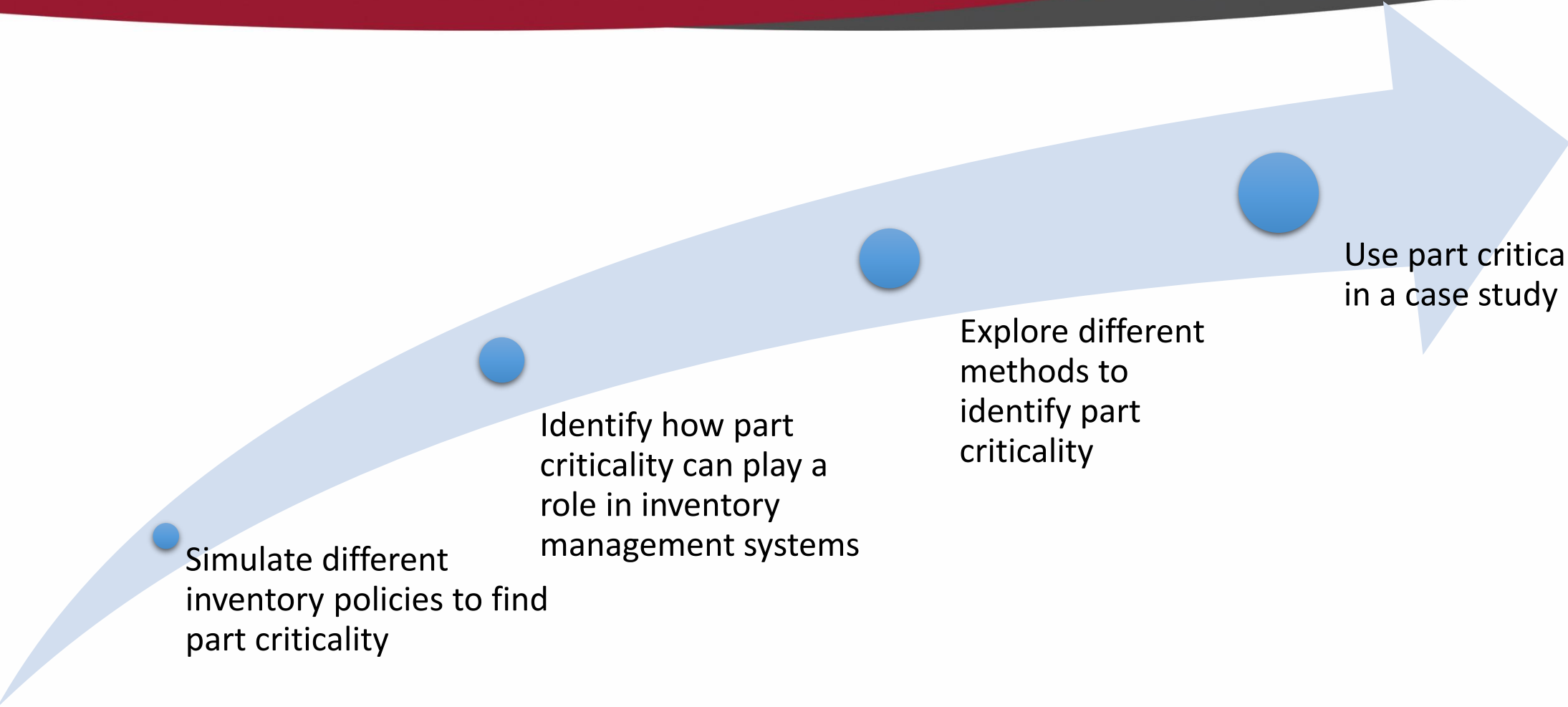
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Aerospace Innovation and Research

PLM Connection

- Helps create **better data transfer**
 - Product structure needs to be in certain format for algorithm to work
- Part criticality can be applied to other times in the product life cycle
 - Which parts break most often/need replacements
 - Similar to **spare parts criticality**
- Increased **demand on suppliers for performance and quality**
 - Important to choose the best possible supplier for timing and quality



NEXT STEPS



Simulate different inventory policies to find part criticality

Identify how part criticality can play a role in inventory management systems

Explore different methods to identify part criticality

Use part criticality in a case study



Thank you

Thank you for coming and your
attention



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