

Manufacturing Cost Prediction in the Presence of Categorical and Numeric Design Attributes


By

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MANUFACTURER'S COST ESTIMATION STRATEGY




The expected cost is
\$ per item.

Step 1. Before Manufacturing



The cost is
\$\$ per item now.

Step 2. During Manufacturing



We cannot make it
less than \$\$\$\$ per
item!

Step 3. After Manufacturing

Critical Questions:

- What is the negotiation power over the underlying product price? Is the expected cost accurate?
- Is it possible to know the cost of a new and unique design before it is actually manufactured?

- When manufacturing a new unique design, the focal point is to establish a price which maximizes customer value while being profitable.
- Since an irreversible and large amount of capital is tied up in production elements, estimating manufacturing costs accurately is critical.
- Final decisions about the product price should be based on analytical approaches, instead of intuitive expectations.

- “Cost plus pricing” or “Cost based pricing”
- Poorly established product prices that are a function of product cost may cause two unfavorable consequences:
 - (1) A potential loss of profit due to the gap between the expected cost and the actual cost
 - (2) A loss of customers and goodwill due to higher prices than necessary



Manufacturer A

\$5 /piece



Manufacturer B

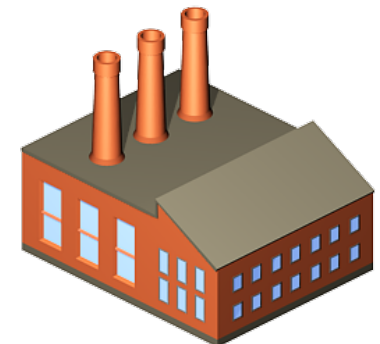
\$4.50 /piece



Tubular Cable Lugs



Customer



Manufacturer C

\$4 /piece

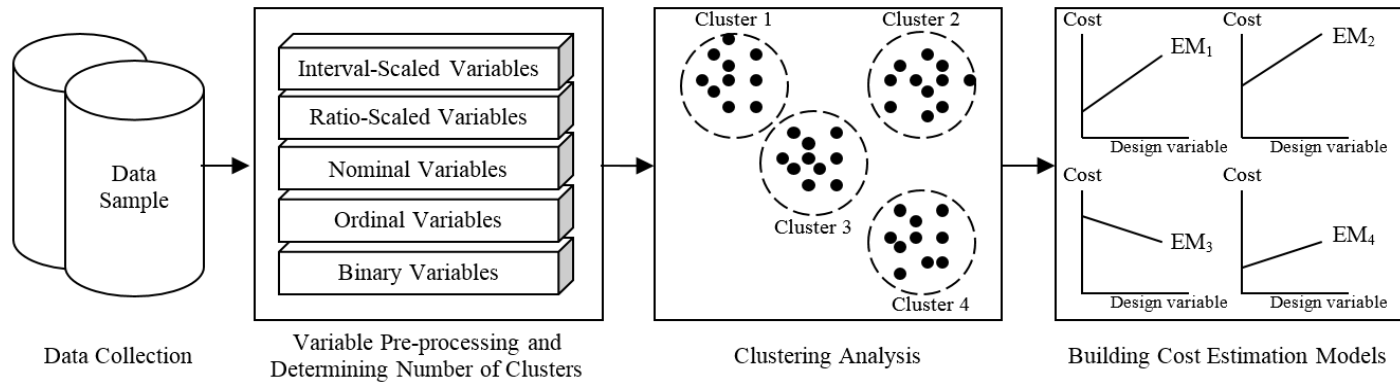
- Many cases, costs are estimated based on primitive intuitive approaches that are far from reality and accuracy.
- Making parametrical distribution assumptions for design attributes can be arbitrary.
- Over a diverse product family, establishing only a single accurate estimation model is challenging and doubtful.

- We would like to predict the manufacturing cost of a product quickly and accurately.
- We investigate ways of using clustering methods to predict the manufacturing cost of products in the presence of complex numeric and categorical design attributes.
- The accuracy of the methodology is assessed in comparison to a traditional approach, a polynomial regression model in absence of a clustering approach.

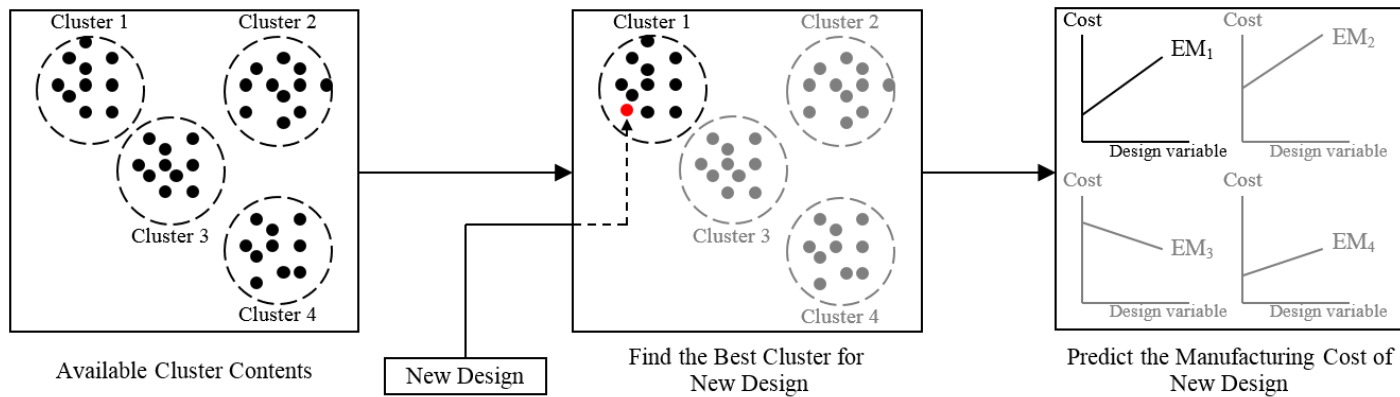
- To accurately and quickly predict the cost of a particular product before it is manufactured
- To deploy clustering techniques to achieve improved accuracy in the prediction
- To find appropriate number of clusters for a given case and series of products

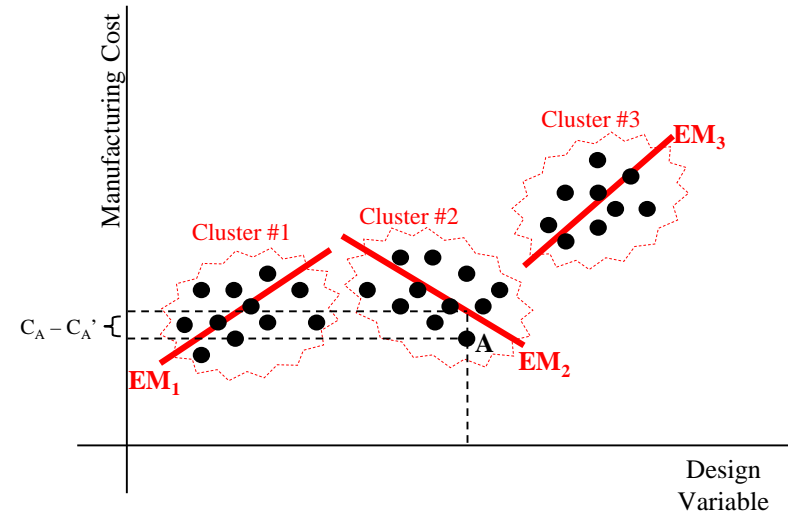
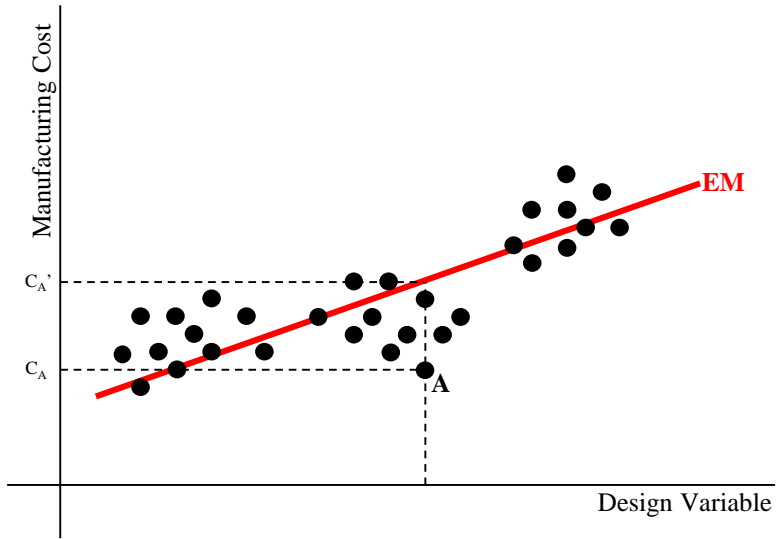
- First to introduce a manufacturing costs estimation approach for mixed categorical and numeric variables using clustering methods
- Implemented a simple heuristic to determine the appropriate number of clusters for mixed data when there is no prior knowledge about the number of product groups

1



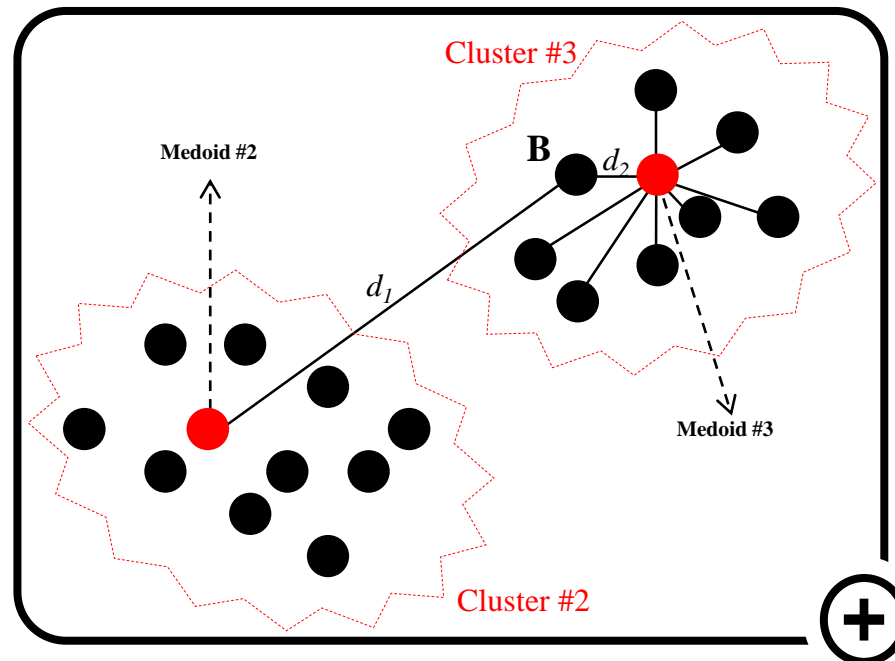
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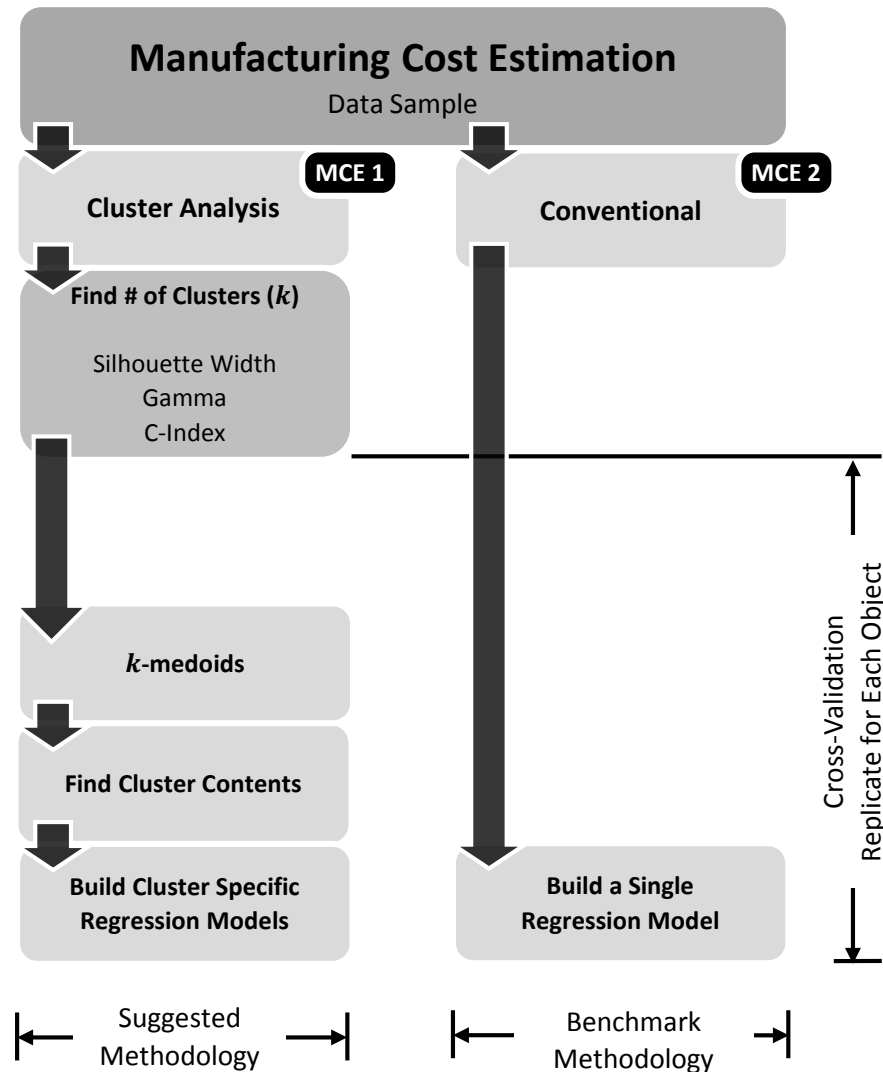


- k -medoids

- Operates on a dissimilarity matrix
- No randomness: Initial solution (BUILD), Moves (SWAP)
- Handles outliers



- Top 6 performing indices (Milligan and Cooper):
 - Calinski and Harabasz's PSF
 - Duda and Hart's $J_e(2)/J_e(1)$ or PST2
 - *Dalrymple-Alford's C-index
 - *Baker and Hubert's Gamma
 - Beale's F-ratio
 - Sarle's CCC
- *Rousseeuw's average silhouette width
- Consensus among Gamma (local peaks), silhouette width (local peaks & > 0.5), C-index (local troughs)

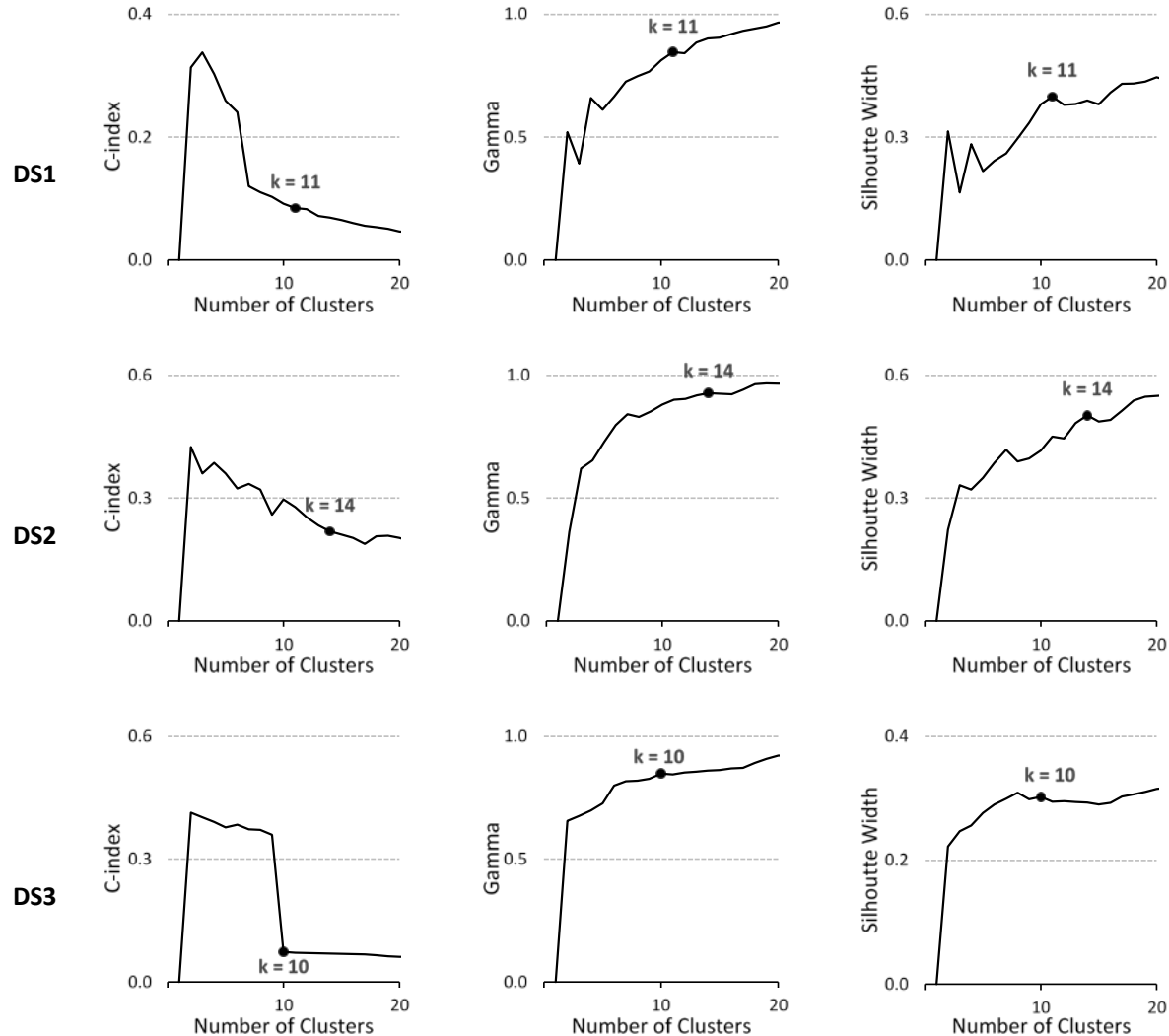


- Electromagnetic and lightning protection parts manufacturer
 - **DS1** Tubular cable lugs: 12 variables
 - **DS2** Air rods: 10 variables



- Plastic kitchen and household products manufacturer
 - **DS3** Plastic parts: 51 variables





MARE

	MCE 1	MCE 2
DS 1	4.98%	49.82%
DS 2	5.81%	15.42%
DS 3	12.39%	33.83%

Min ARE

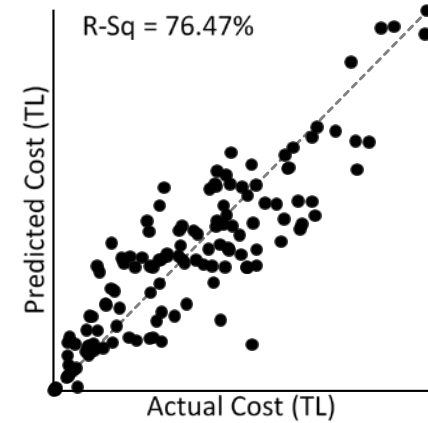
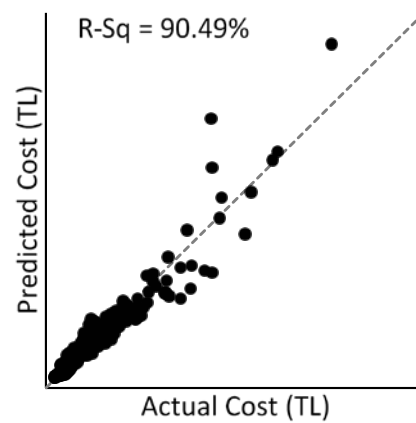
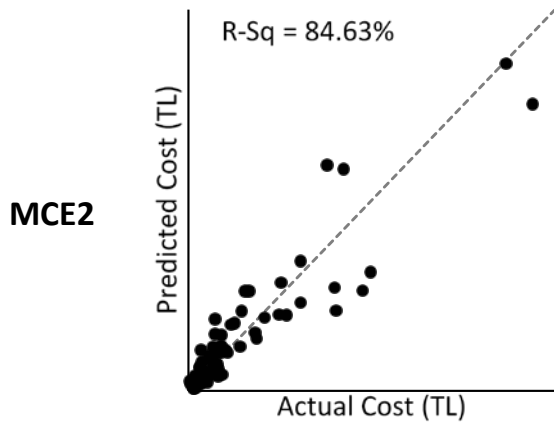
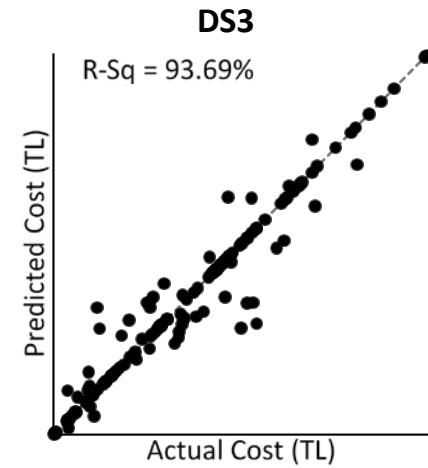
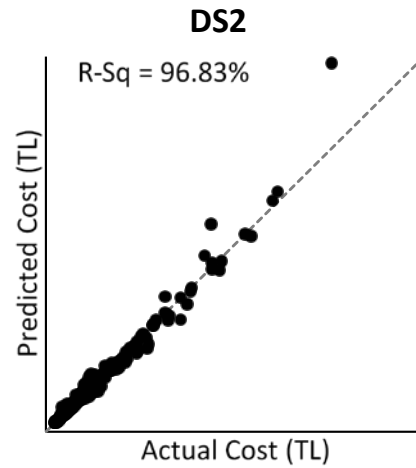
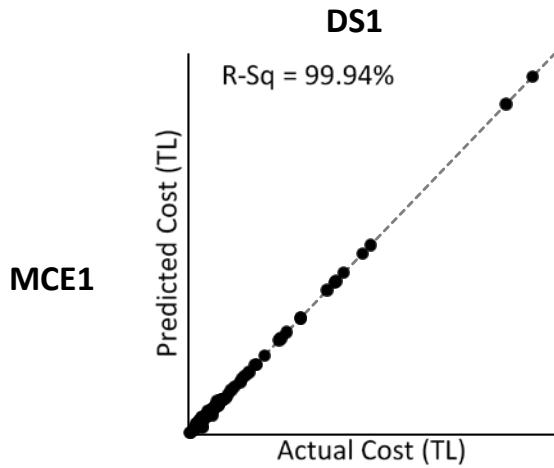
	MCE 1	MCE 2
DS 1	0.00%	0.00%
DS 2	0.00%	0.00%
DS 3	0.00%	0.00%

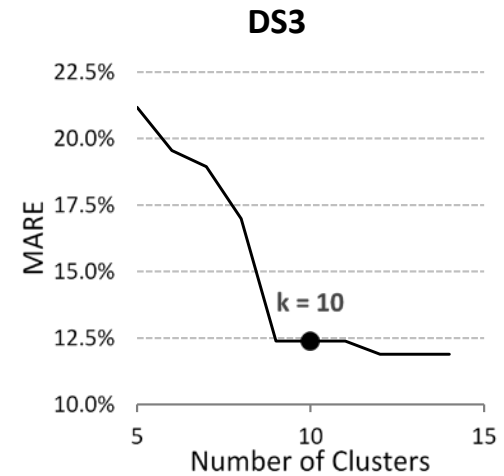
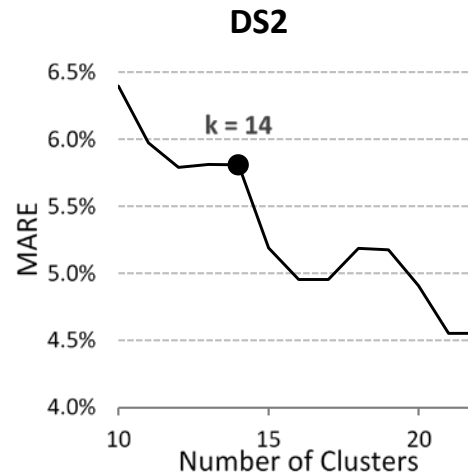
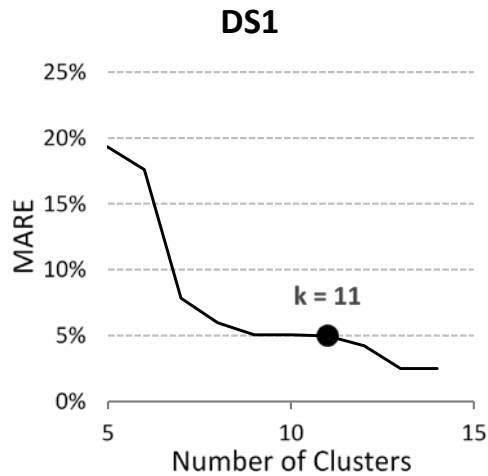
RMSE

	MCE 1	MCE 2
DS 1	8.86%	140.26%
DS 2	355.72%	615.92%
DS 3	17.71%	34.20%

Max ARE

	MCE 1	MCE 2
DS 1	46.67%	429.52%
DS 2	56.04%	64.36%
DS 3	203.54%	233.79%





- We investigated ways of using clustering methods to predict the manufacturing cost of a product without actually manufacturing it.
- The accuracy of the methodology is assessed in comparison to a simple regression model with the absence of clustering approaches.
- The main concern is to predict the manufacturing cost of a product without dealing with arbitrary assignments of statistical distributions to cost related attributes.

- In real production systems often a variety of products are being manufactured under a single facility roof.
- Over a diverse product family, establishing only a simple accurate estimation model is challenging and even questionable.
- This motivated us grouping products according to their design features, common manufacturing operations or some other factors by dividing the whole database of products into neighborhoods.
- Then for each group of products (clusters), a cost estimation model is developed to predict the manufacturing cost of a new product with using the cluster specific model.

- Developing a comprehensive similarity measure that demonstrates high inter-cluster variability while being able to handle mixed categorical and numeric design attributes.
- A deterministic model such as a mixed integer programming model can be implemented to obtain the optimal cluster results.
- Information gain criterion can be considered when deciding on the inclusion of a candidate predictor (design attribute) in the cost estimation model.

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