Using Sequential CUSUM Permutation Tests to Identify Process Shifts

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Outline

- Definitions
 - Detection methods
 - -CUSUM estimator
- Sequential Permutation Method
- Simulation Results
- Simulated Case Study JMP Demo
- Conclusions





Methods of Detection

- Visual (Simple but Subjective)
 - -Raw data
 - -CUSUM
 - -EWMA
- Analytical (Complicated but Objective)
 - Change-Point estimators; i.e. CUSUM, EWMA
 - Mathematical Modeling; i.e. MLE, SSE
 - Permutation Tests (Proposed Method)





Types of Variation

 Common Causes – natural (random) variations that are part of a stable process
 Machine vibration
 Temperature, humidity, electrical current fluctuations
 Slight variation in raw materials

 Special Causes – unnatural (non-random) variations that are not part of a stable process
 Batch of defective raw material
 Faulty set-up
 Human error
 Incorrect recipe





Cumulative Sum Control Chart

CUSUM: cumulative sum of deviations from average

$$C_{t}^{+} = \max\left[0, \overline{y}_{t} - (\mu_{0} + K) + C_{t-1}^{+}\right]$$

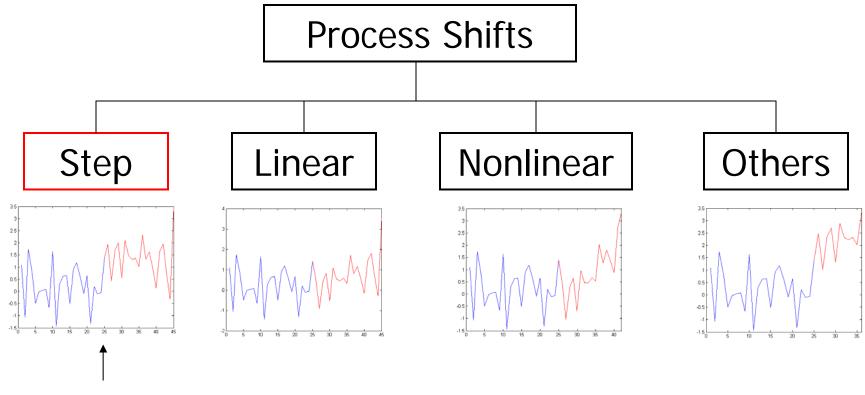
 C_t^{\pm} = accumulate d deviations above/belo w μ_0

- A bit more difficult to set up
- More difficult to understand
- Very effective when subgroup size n=1
- Very good for detecting small shifts
- Change-point detection capability
- Less sensitive to autocorrelation





Types of Process Shifts

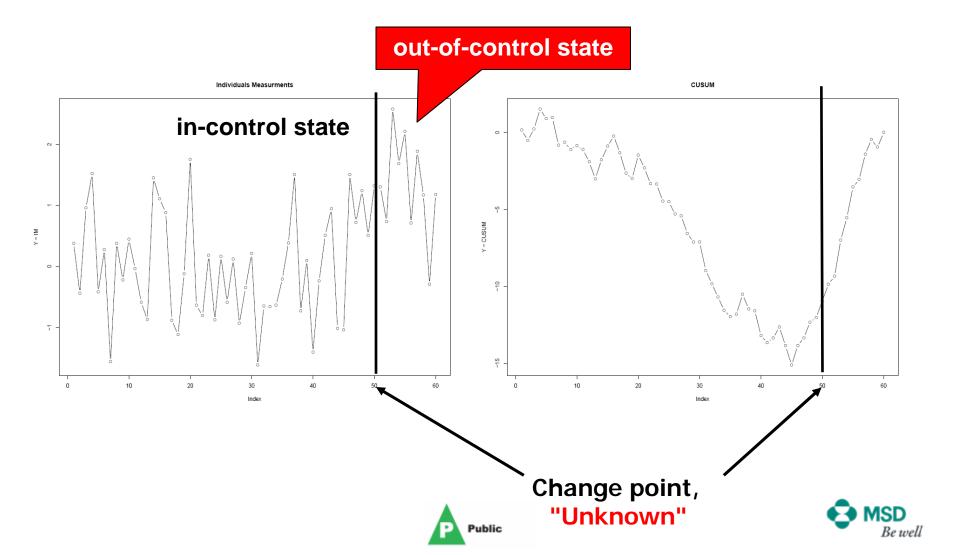


Our Example Today





Process Model



Sequential Permutation Test

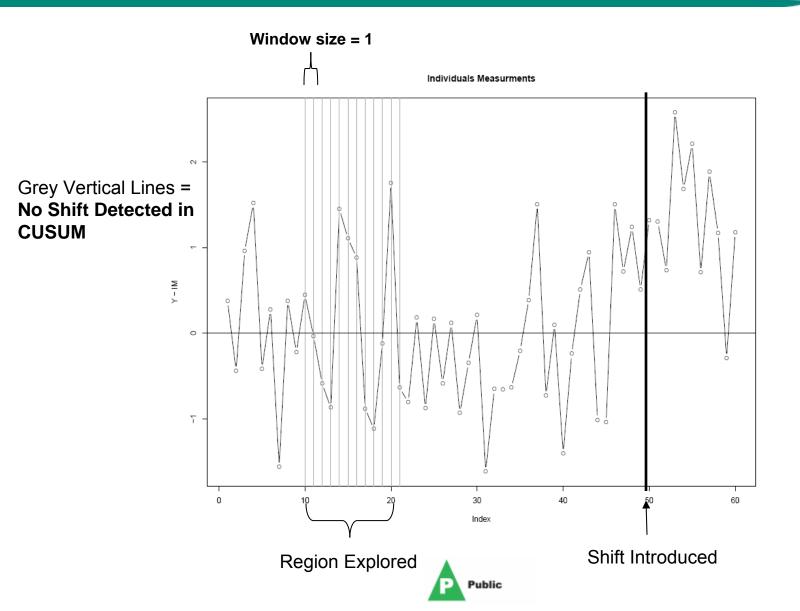
Sequential Test

- In order to determine where the change point may have occurred choose a region within the time series where you wish to determine if a shift in the process occurred
- Between adjacent points (window), within this region, determine the value difference between your CUSUM statistic
- Randomly permute the CUSUM, each time calculating the value difference for the corresponding points
- Use the percentile method to determine the proportion of times your calculated difference was greater than the calculated differences in the random permutations
 - If >95%, you can conclude that a shift may have occurred





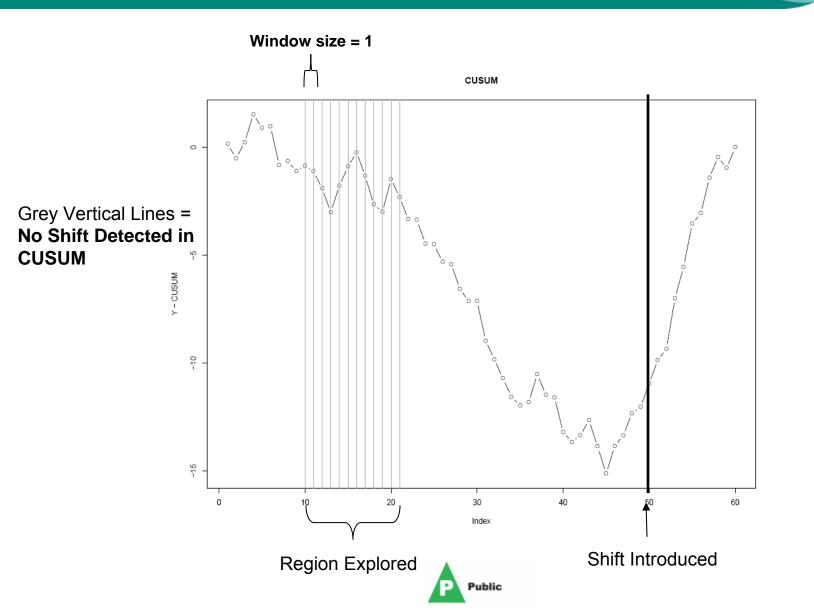
Stable Area Individuals Measurement Chart



MSD

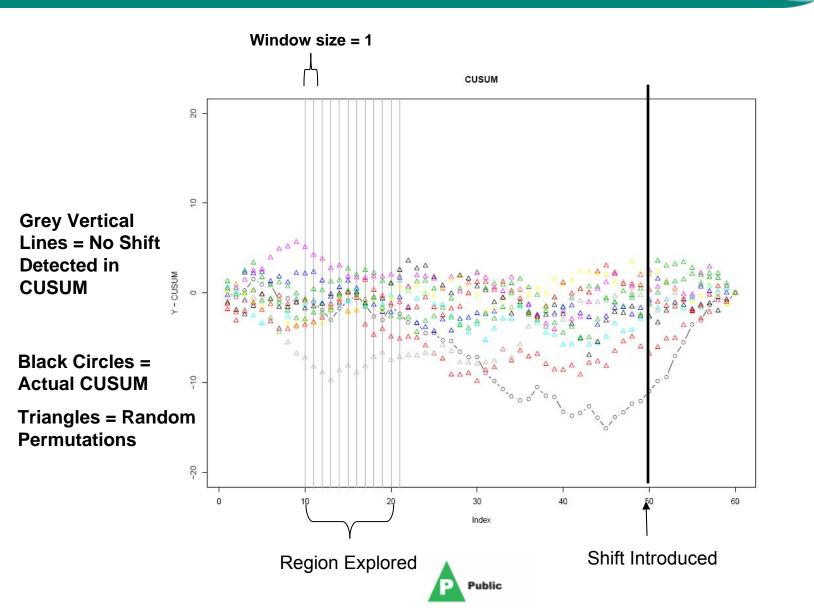
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Stable Area CUSUM Chart





CUSUM Chart with Random Permutations





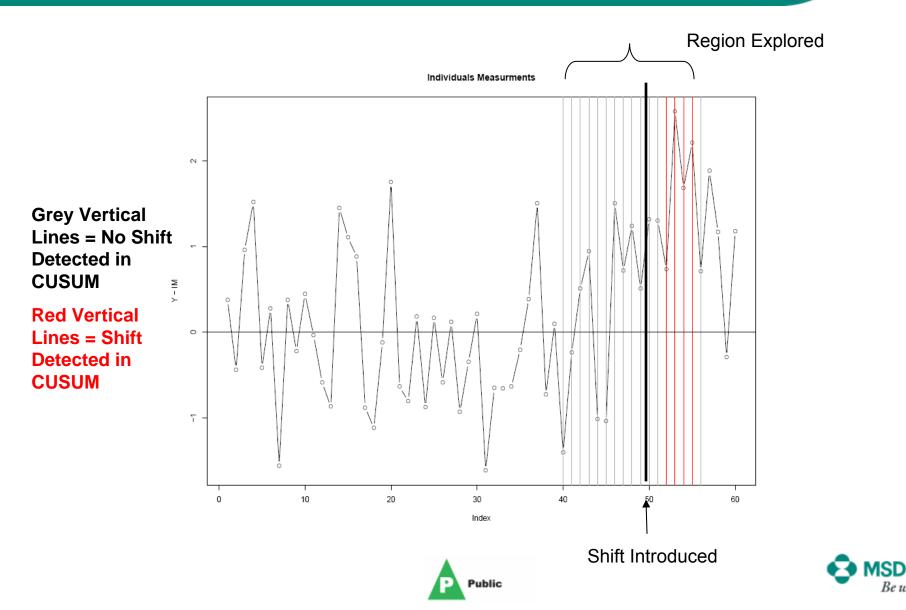
Distribution of Randomly Permuted and Actual Delta's Calculated from CUSUM

Windov	w							
X	Y	Actual Delta	Random Delta 1	Random Delta 2		Random Delta n	I	
40	45	6	1	1		4		
41	46	1	2	5		2		
42	47	2	3	2		7		
43	48	1	5	3		0		
44	49	5	5	4		4		
45	50	2	3	0	•••	6		
46	51	10	2	4		1		
47	52	4	2	6		6		
48	53	4	5	6		4		
49	54	5	2	1		1		
50	55	7	6	5		4		
					-		Deltas From Random Permutations	∆ Actual Delta



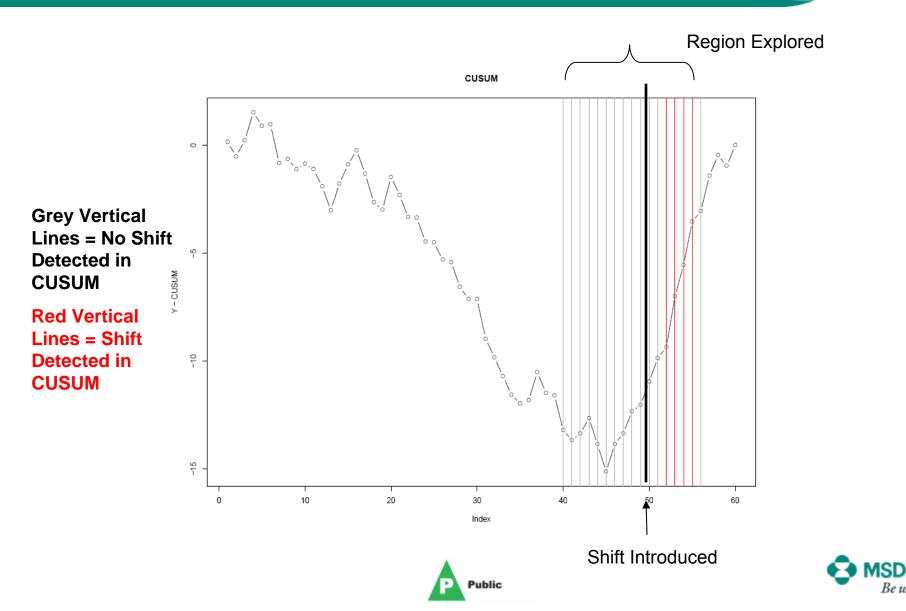


Shift Area Individuals Measurement Chart



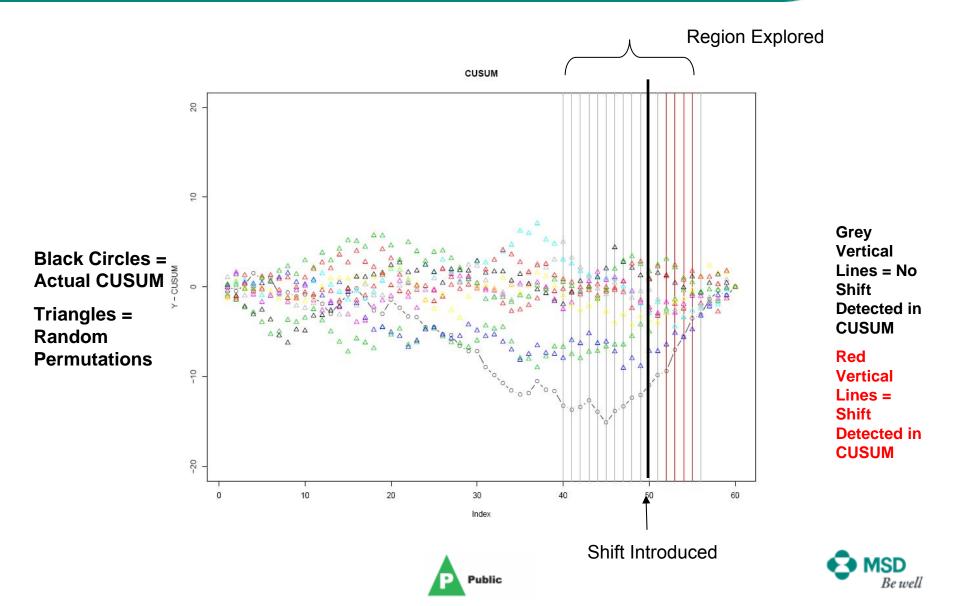
Be well

Shift Area CUSUM Chart



Be well

Shift Area CUSUM w/ 10 Random Permutations



Sensitivity Analysis

- 1000 Simulation Runs
- 1000 Permutations
- Window Size: 3 and 5
- Test @ S/N: 0, 1, 1.5 and 2





Signal-to-Noise 0 Window Size 5 1.0 -Proportion of Times Shift Detected Proportion of Times Shift Detected 0.0 15 Signal-to-Noise 0 Window Size 3 1.0 -Proportion of Times Shift Detected

10

Seq

5

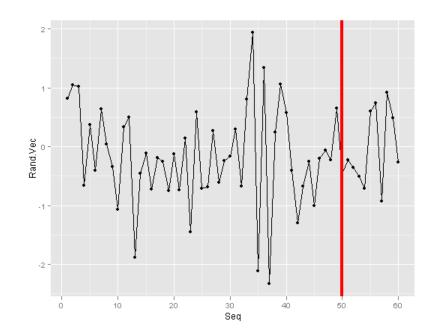
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Public

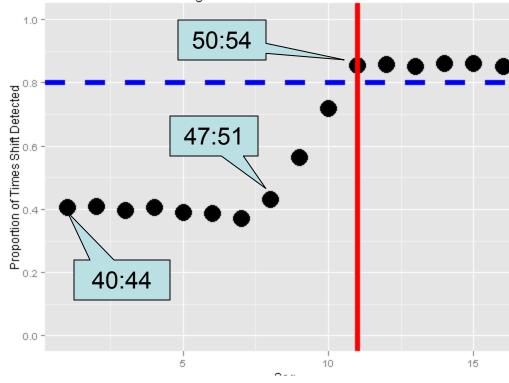
0.0 -

S/N=0, Wind=5





Signal-to-Noise 1 Window Size 5



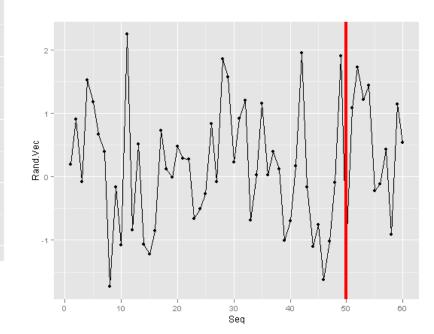
Signal-to-Noise 1 Window Size 3

5

10

Seq

S/N=1, Wind=5

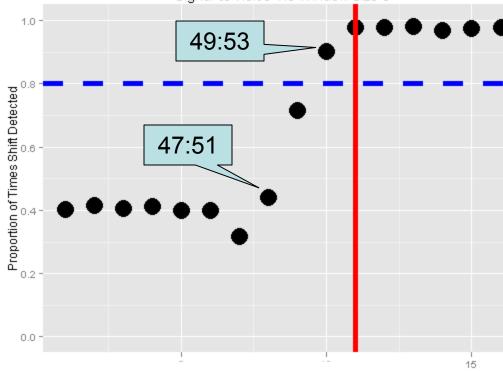




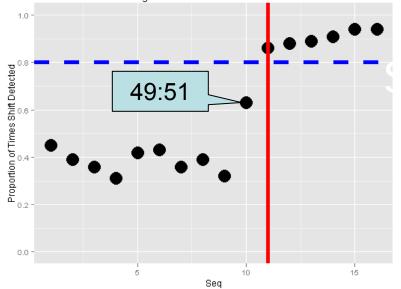


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Signal-to-Noise 1.5 Window Size 5

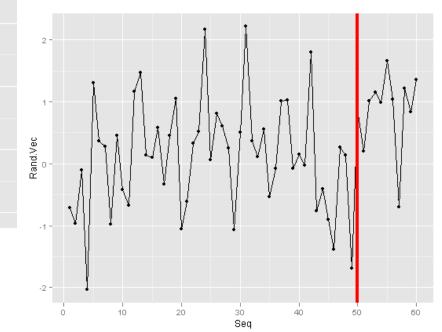


Signal-to-Noise 1.5 Window Size 3



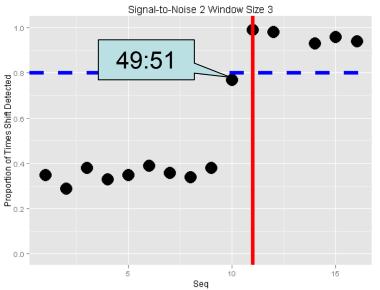
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S/N=1.5, Wind=5





Signal-to-Noise 2 Window Size 5 S/N=2, Wind=5 1.0 -48:52 Proportion of Times Shift Detected 4 2 Rand.Vec 0 0.0 -10 15 5 ~ -2





50

60

1 40

30 Seq

20

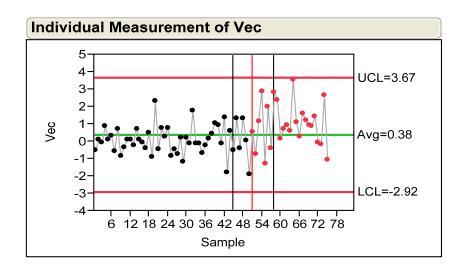
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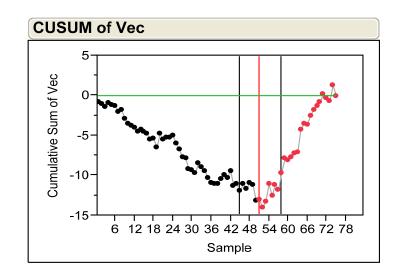
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Integration of JMP[®] and R

Sequential Permutation Test Simulated Process





Assumption: Normal distributed data with no auto-correlation





Integration of JMP[®] and R

Sequential Permutation Test GUI

Image: Distribution Image: Distribution		nalyze Graph Tools View Window Help	🕴 Select Columns
45 45 -0.522283351		yx Fit Y by X Matched Pairs Fit Model Modeling CUSUM Sequential Permutation Test Multivariate Methods Reliability and Survival 43	Select Columns Cast Selected Columns into Roles Seq Time Series Vec Remove Permutations 1000 Start Point 45 End Point 48
46 46 1.322230956 OK Cancel			
47 47 -0.363440327	47	47 -0.363440327	

Window Size = 3

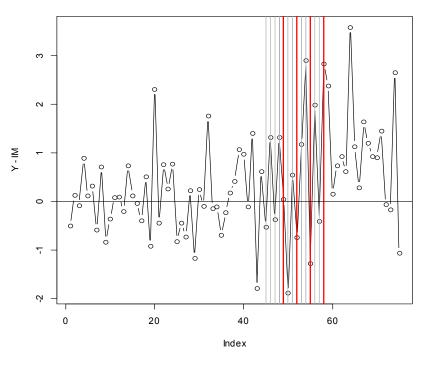




Integration of JMP[®] and R

Sequential Permutation Test Output & Results w/ Actual Output

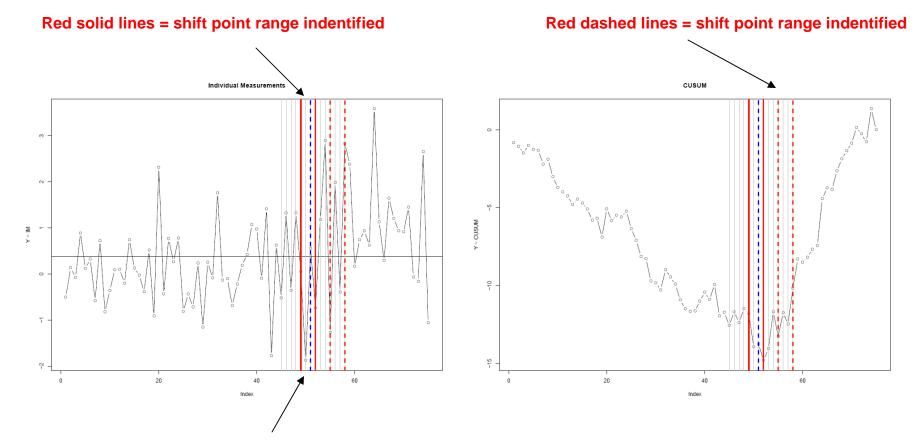
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		•	1	74.5	25.5	-	45	48	
			2	51.2	48.8	No	46	49	
			3	81.3	18.7	No	47	50	
💌 Columns (5/0)			4	93.1	6.9	No	48	51	
⊿ Did.Shift	_1		5	97.3	2.7	Yes	49	52	
🚄 No.Shift			6	50.5	49.5	No	50	53	
📙 Significant			7	87.7	12.3	No	51	54	
A X			8	80.3	19.7	No	52	55	
🚄 Y			9	91.5	8.5	No	53	56	
			10	66.5	33.5	No	54	57	
	_		11	96.3	3.7	Yes	55	58	
- Rows									
All rows	11								
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Simulated Case Study Results



Blue dashed line = potential shift





Conclusions

- Change-point problem is general and can be applied in many applications such as degradation curves and linear/non-linear trends, in addition to time series models.
- Another application in manufacturing processes includes detection of the change-point for process variance.
- It is preferred to combine both analytical and visual techniques; in addition to the process expertise; to get accurate results.
- Permutation tests can provide an objective test of where a change-point may have occurred
- Future work:
 - Apply methodology to EWMA
 - Apply to linear trends
 - Impact of auto-correlated data







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