# Phase I Analysis of High Dimensional Data Using Multivariate Control Statistics

ISEN 614: Advanced Quality Control Project
Team no. 2

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# Purpose

- The purpose of this project is to conduct a Phase I analysis on a dataset with 209 characteristics and 552 observations with sample size of 1
- The end goal is to establish control charts with in-control mean and covariance so that future observations can be monitored in Phase II analysis
- To do so, multivariate statistical methods were used to isolate all out-of-control (OOC) data from the in-control (IC) data

# Approach

- Because the physical meaning of the dataset was unknown, PCA was performed on the covariance and the correlation matrix
- The number of PCs to retain had to account for close to 80% of the total variation for substantial representation of data
- Since the total number of observations equal 552, the $ARL_0$ had to be substantially less in order to observe sufficient OOC points
- Multiple iterations of the $T^2$ and m-CUSUM control charts were needed in-conjunction in order to ensure both control charts had all in-control data

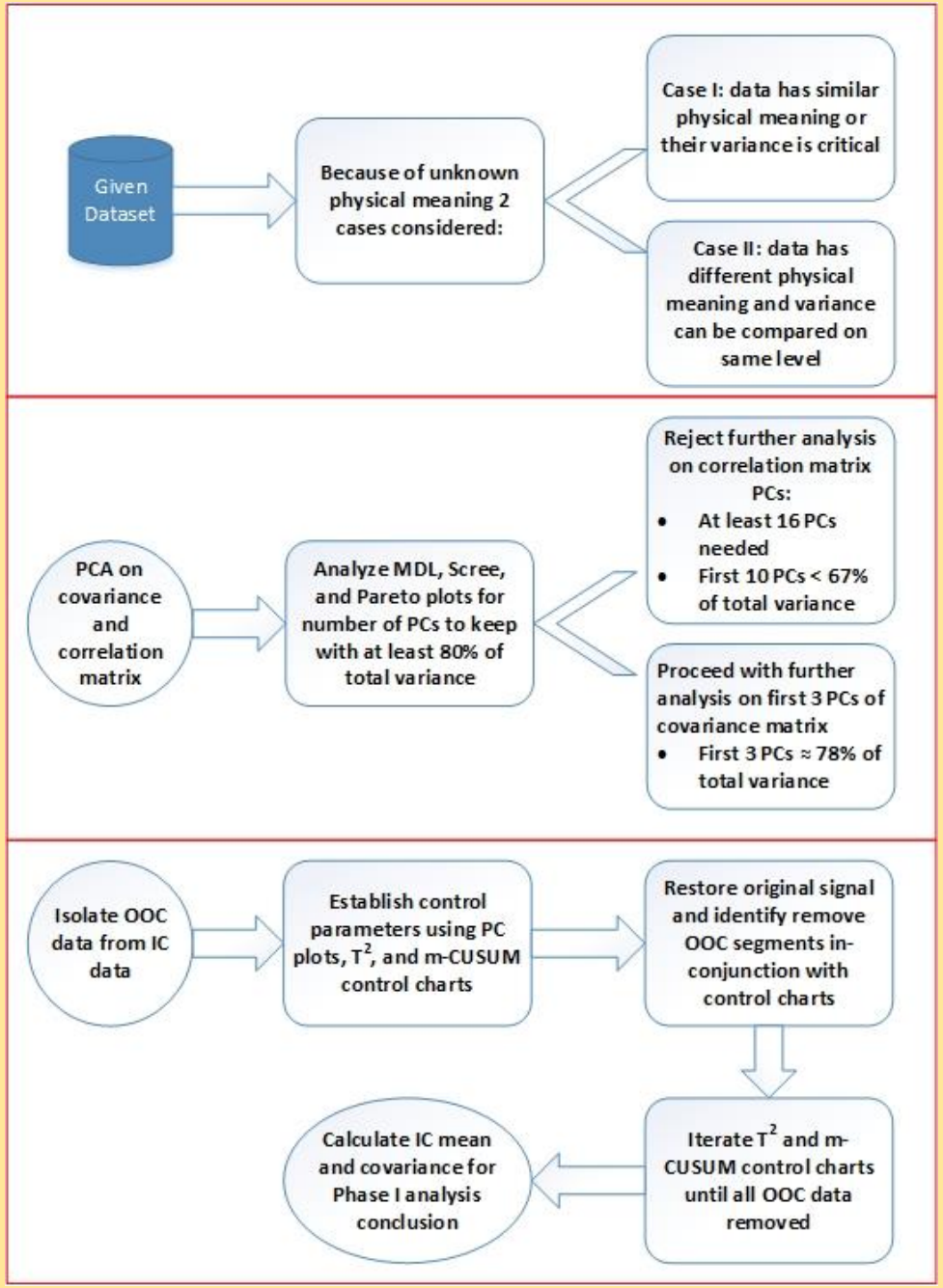


**Figure 1: Pictorial Summary of Approach**

# Principle Component Analysis of Correlation Matrix

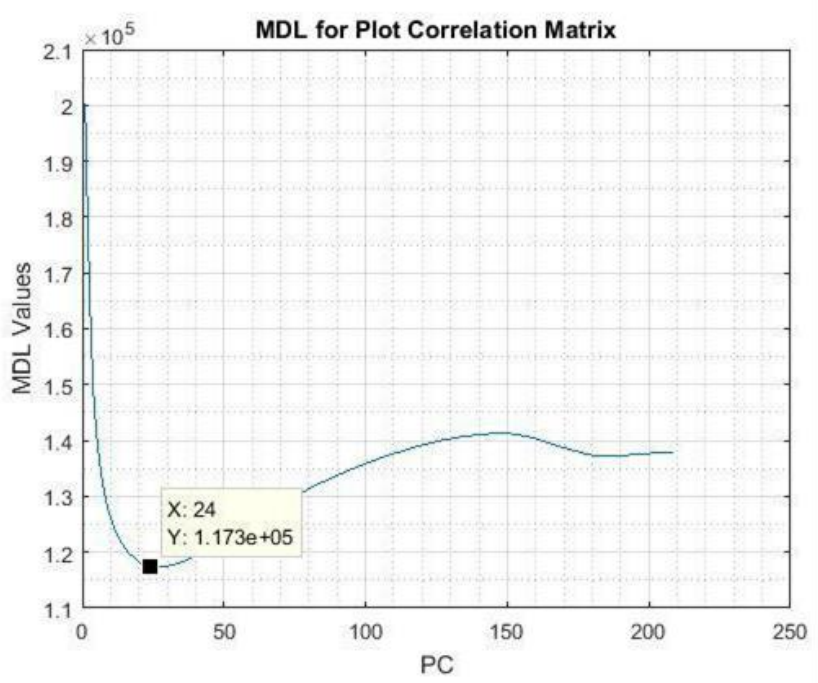


**Figure 2: MDL Plot for PCA on Correlation Matrix**

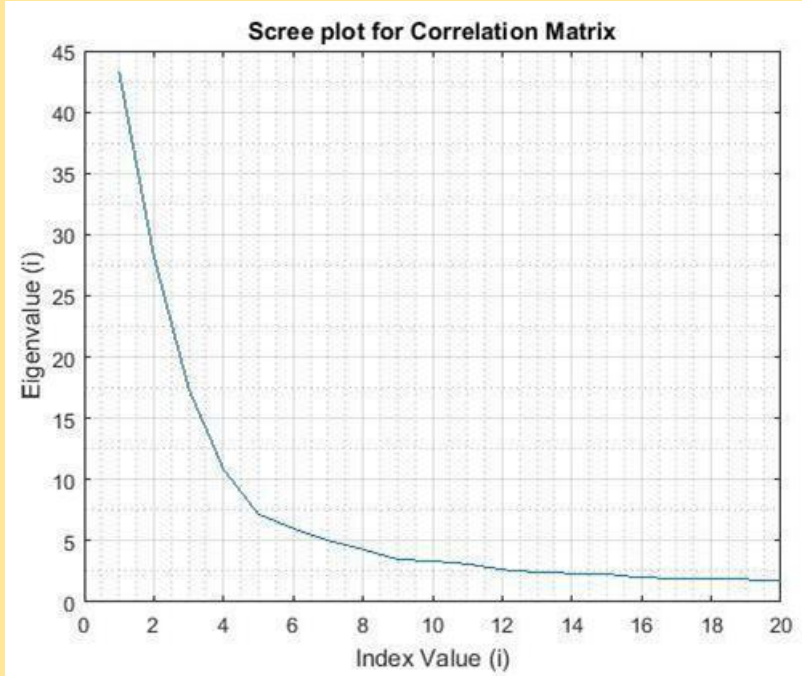


**Figure 3: Scree Plot for PCA on Correlation Matrix**

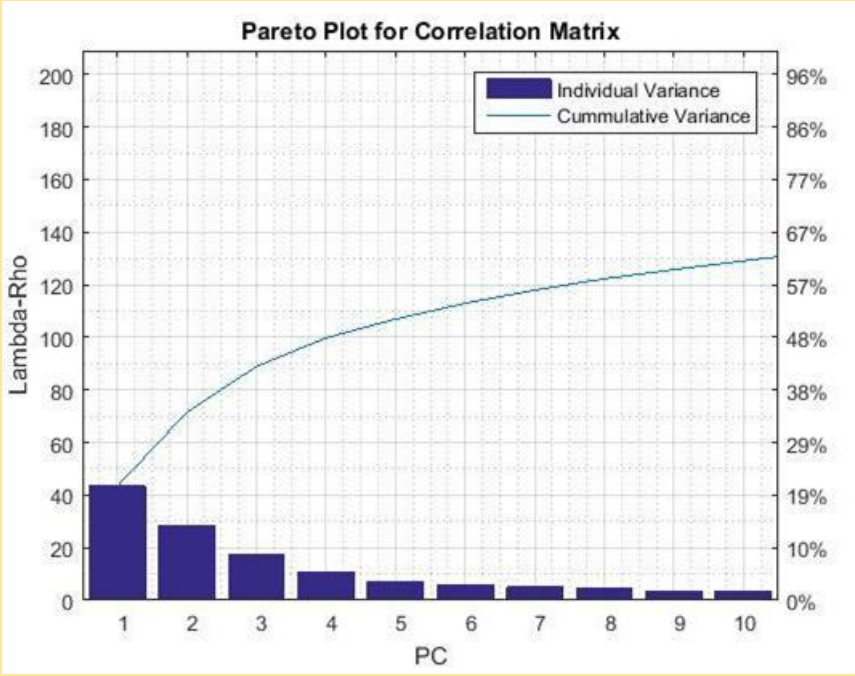


**Figure 4: Pareto Plot for PCA on Correlation Matrix**

- MDL plot retains 24 PCs, which is too many
- Scree plot begins to level only after 16 PCs, but retaining 16 PCs is still too many
- Pareto plot shows 10 PCs contribute less than 67% of the total variance
- Because a total variance of around 80% is not reached with a small number of PCs, further analysis with the PCs for the correlation matrix was forgone

# Principle Component Analysis of Covariance Matrix

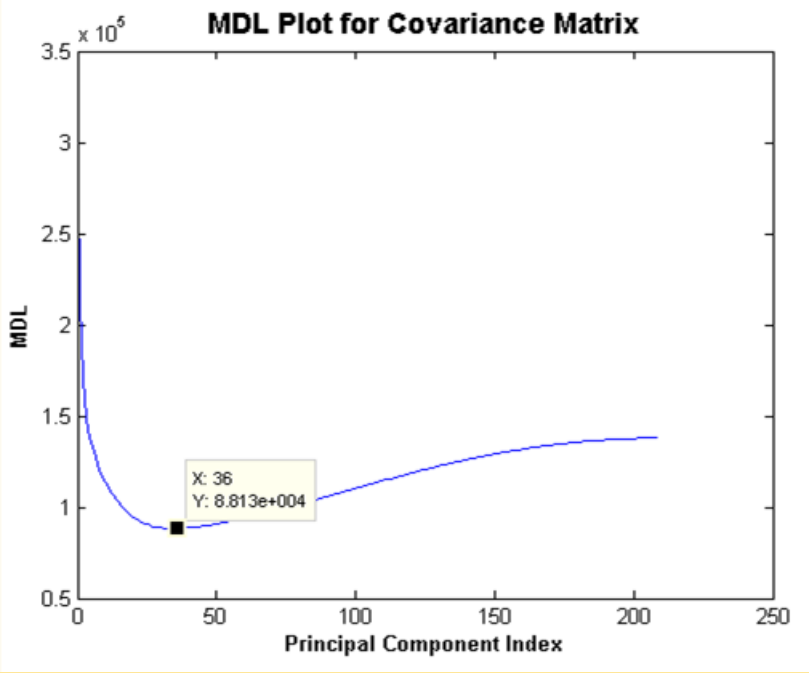


Figure 5: MDL Plot for PCA on Covariance Matrix

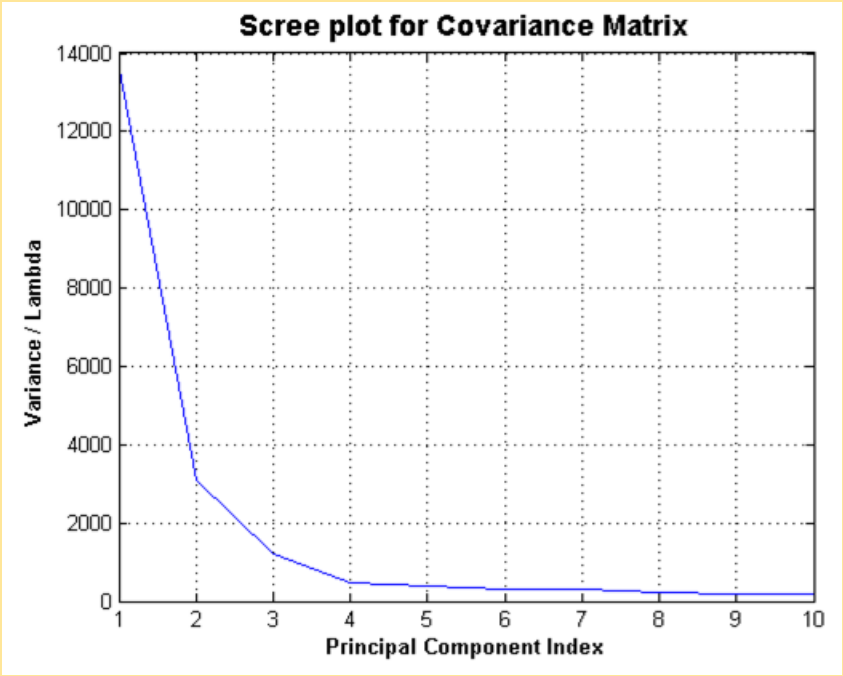


Figure 6: Scree Plot for PCA on Covariance Matrix

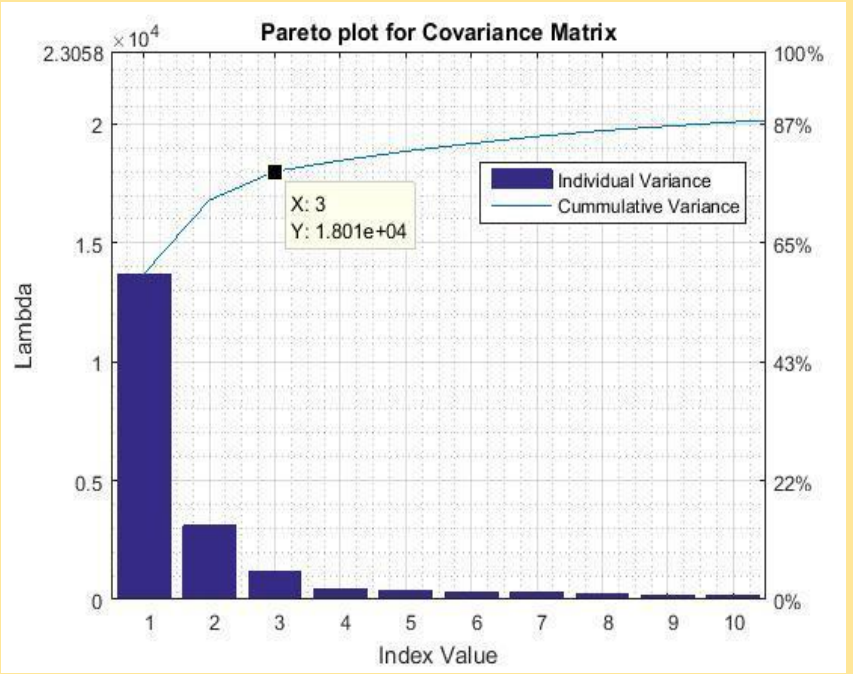


Figure 7: Pareto Plot for PCA on Covariance Matrix

- MDL plot retains 36 PCs, which is too many
- Scree and Pareto plots show 3 PCs have a total variance of 78%, which approximately meets our criteria of a total variance of 80%
- The first 3 PCs of the covariance matrix were chosen for further analysis

# Establishing Control Chart Parameters: m-CUSUM

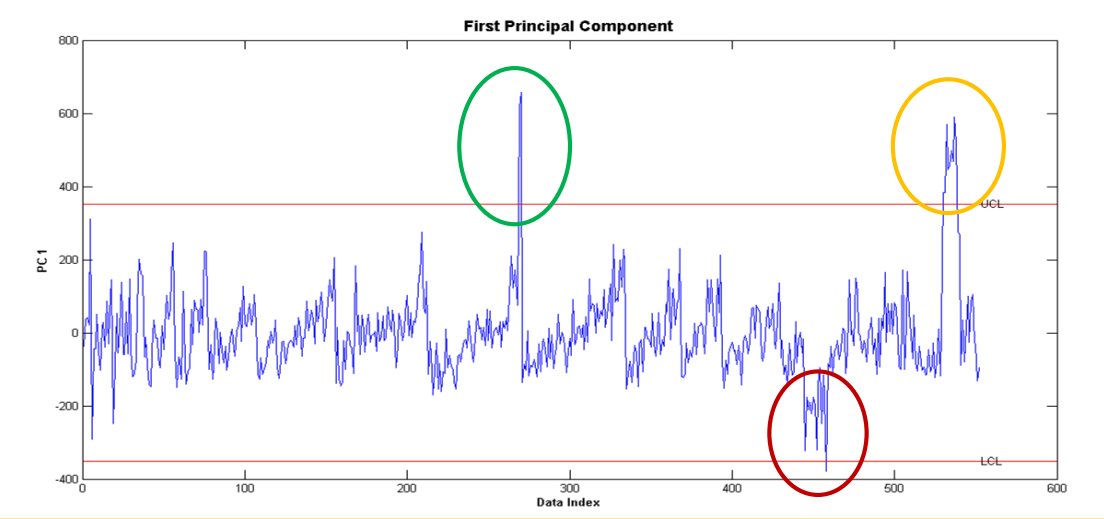


**Figure 8: PC1 for PCA on Covariance Matrix**

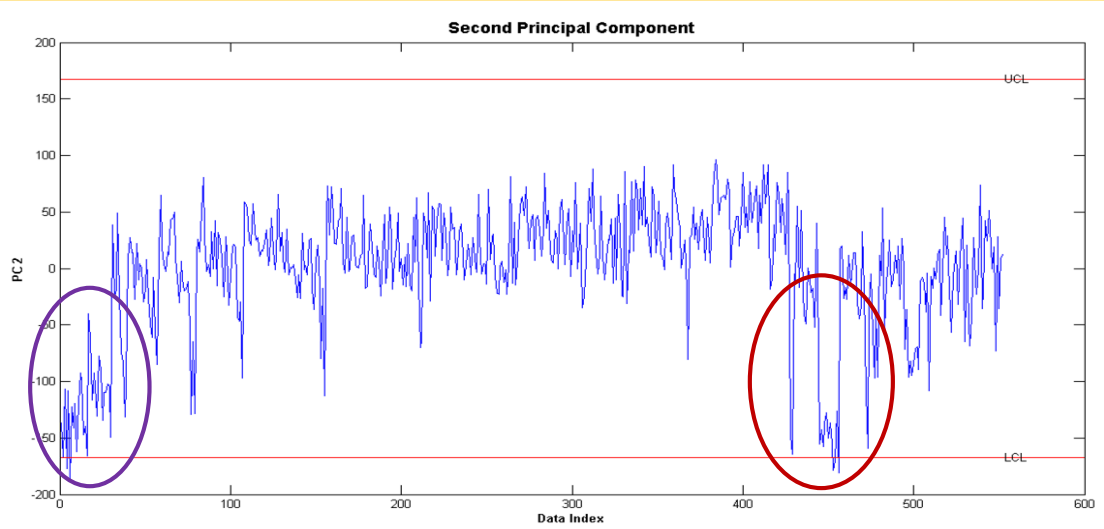


**Figure 9: PC2 for PCA on Covariance Matrix**

- OOC data was observed on PC1 and PC2, but not on PC3
- The first 38 points observed were OOC on PC2 and on the m-CUSUM for the 3-sigma mean shift
- All three charts signaled for data points from 500 to 552
- Because the m-CUSUM closely identifies the same OOC points as the PCs, a **mean shift of 3-sigma** was chosen instead of a mean shift of 2-sigma
- Chosen parameters: **$ARL_0$ = 200** and **UCL = 5.48**

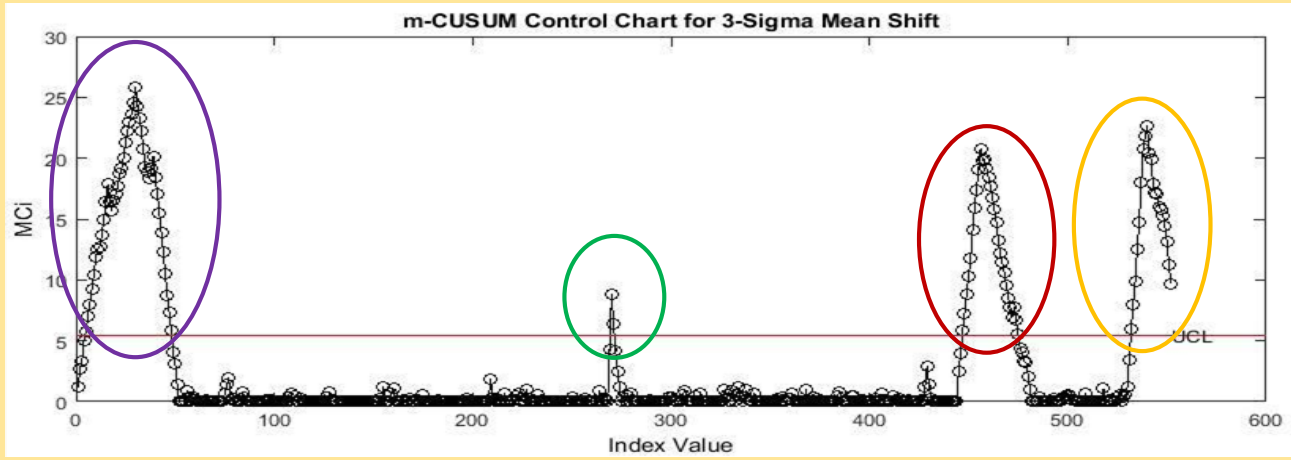


**Figure 10: m-CUSUM for 3-Sigma Mean Shift**

# Establishing Control Chart Parameters: $T^2$

- The $T^2$ chart was able to detect a similar trend because of its sensitivity to large, sustained mean shifts
- An **alpha of 0.005** was used corresponding with an $ARL_0$ of 200

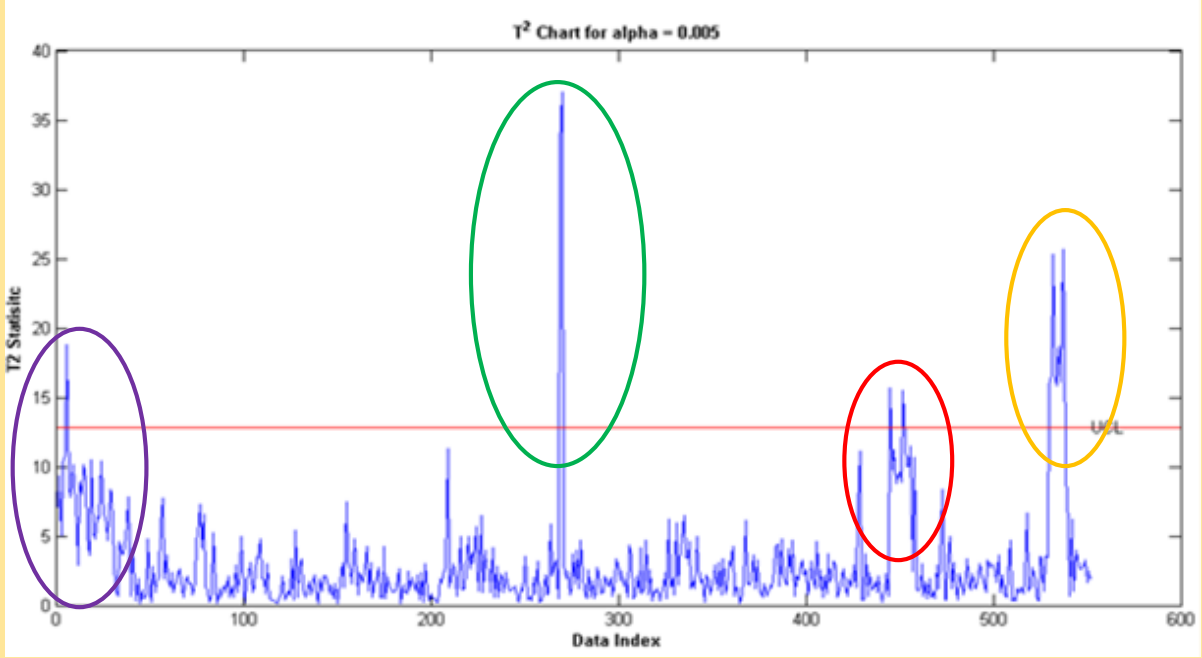


**Figure 11: T2 Control Chart for 3 PCs**

# Restoration of Original Signal Profiles

- Plot shows original profile signal for 3 OOC segments
- The 4th segment shows the profile for the IC data points
- Segment 3 and segment 2 deviate the most from the IC segment

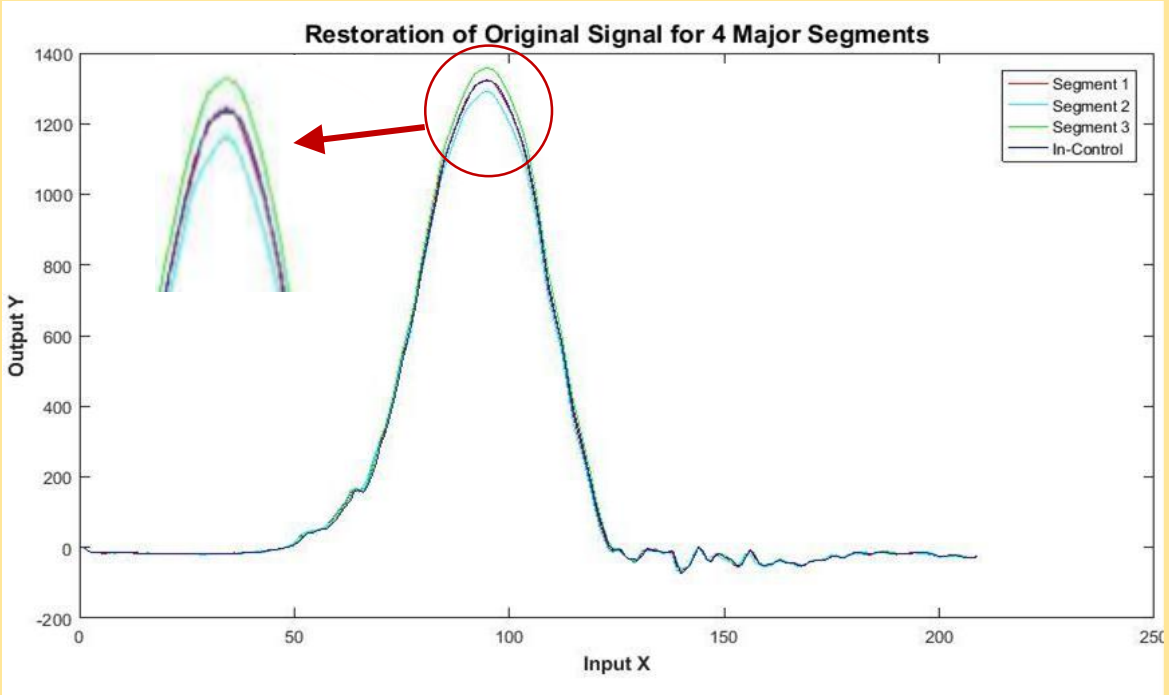


**Figure 12: Restoration of Original Signal Profiles for 4 Major Segments**

# Isolate OOC Data from IC Data: T2 and m-CUSUM Iterations

- The 1st round of $T^2$ was followed by the 1st round of m-CUSUM to establish IC points on both charts
- The 2nd round of $T^2$ showed OOC points, so multiple rounds were needed for both charts to remove all OOC data points

| $T^2$ and m-CUSUM Control Chart Out-of-Control Observations | | | | |
|---|---|---|---|---|
| **Round** | **Control Chart** | **Iteration** | **OOC Points** | **Remaining Observations** |
| One | $T^2$ | 1 | 9 | 458 |
| | | 2 | 7 | 451 |
| | | 3 | 1 | 450 |
| | mCUSUM | 1 | 5 | 445 |
| | | 2 | 1 | 444 |
| Two | $T^2$ | 1 | 3 | 441 |
| | | 2 | 1 | 440 |
| | -CUSUM | 1 | 1 | 439 |
| | | 2 | 1 | 438 |
| Three | $T^2$ | 1 | **0** | **438** |

**Table 1: OOC Points for Control Chart Iterations**

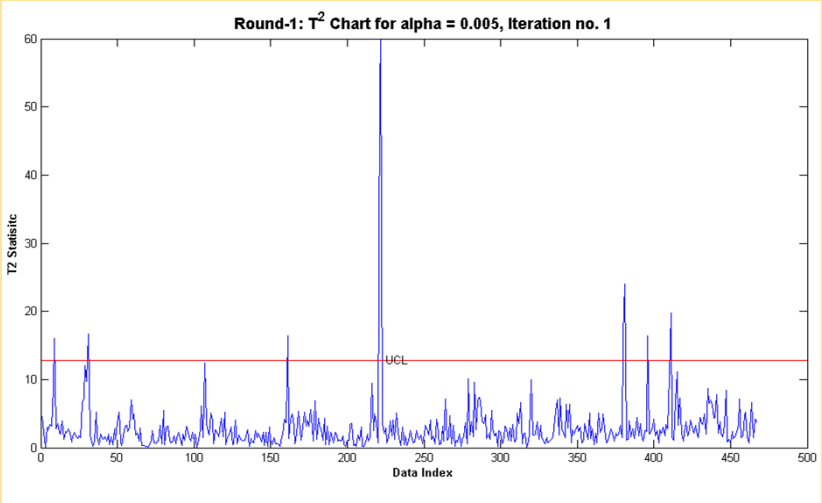


**Figure 13: $T^2$ Round 1 Iteration 1**

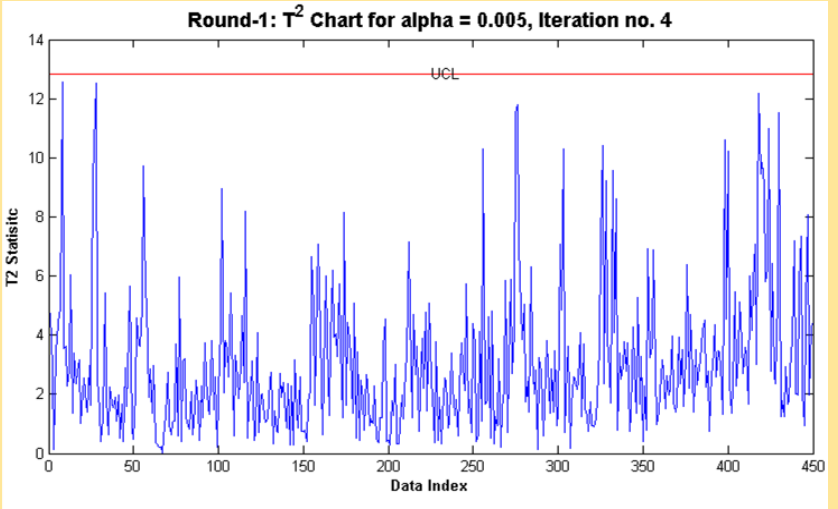


**Figure 14: $T^2$ Round 1 Iteration 4**

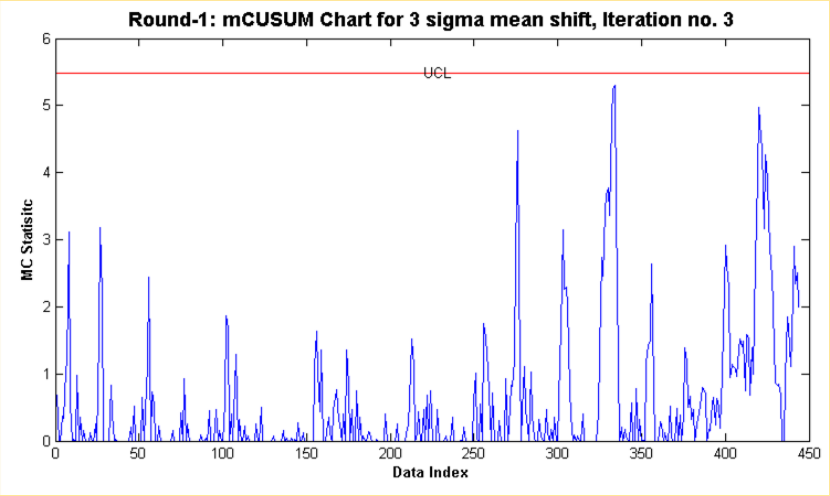


**Figure 15: m-CUSUM Round 1 Iteration 3**

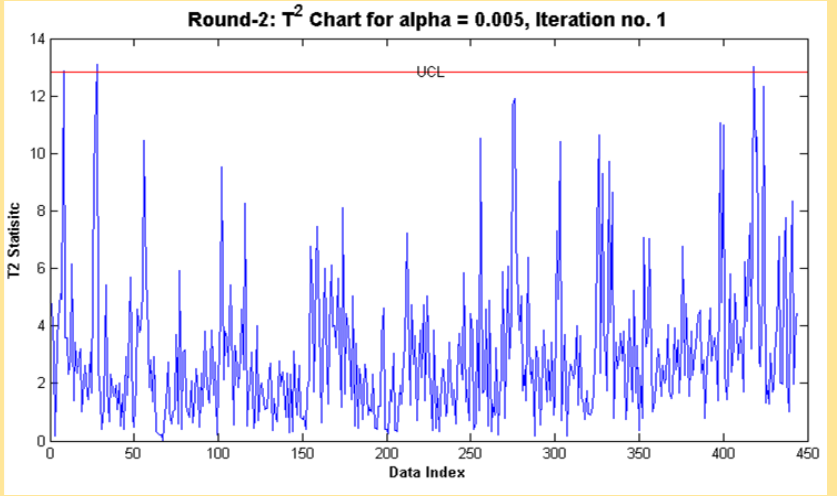


**Figure 16: m-CUSUM Round 1 Iteration 4**

# Results

- After 3 rounds of iterations, all OOC data points were removed
- A total of 438 observation out of the original 552 were found to be in control
- In-control mean: $\begin{bmatrix} 11.27567 \\ 16.49829 \\ 5.62782 \end{bmatrix}$
- In-control covariance: $\begin{bmatrix} 5724.7 & 519.95 & -456.43 \\ 519.95 & 1180.8 & -354.18 \\ -456.43 & -354.18 & 1084.2 \end{bmatrix}$

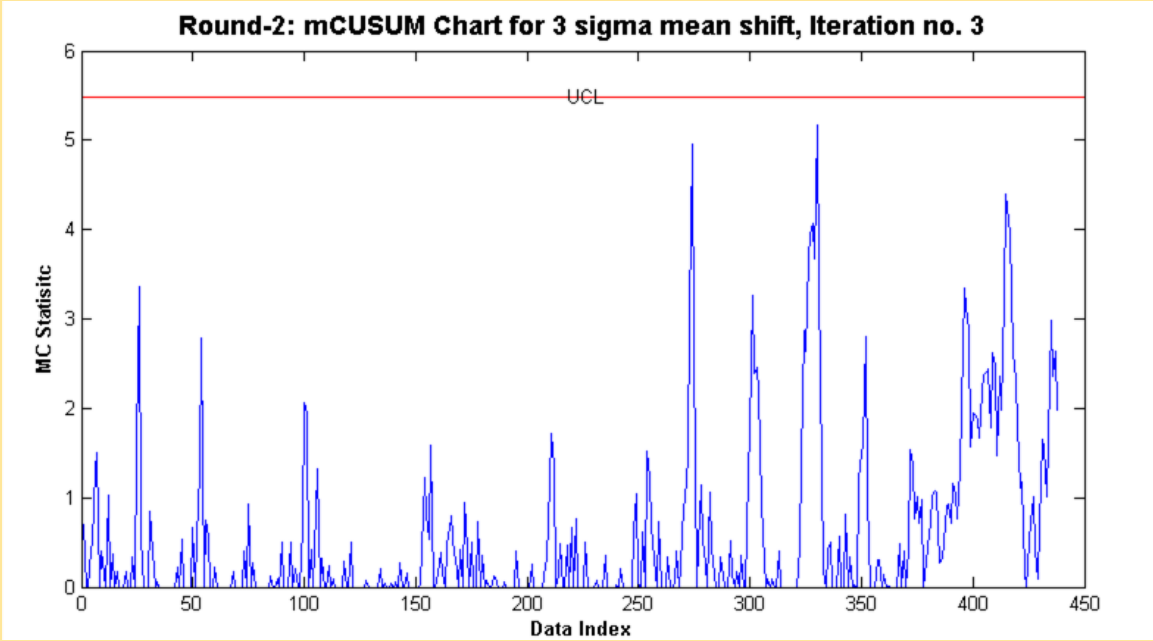


**Figure 17: m-CUSUM Round 2 Iteration 3**

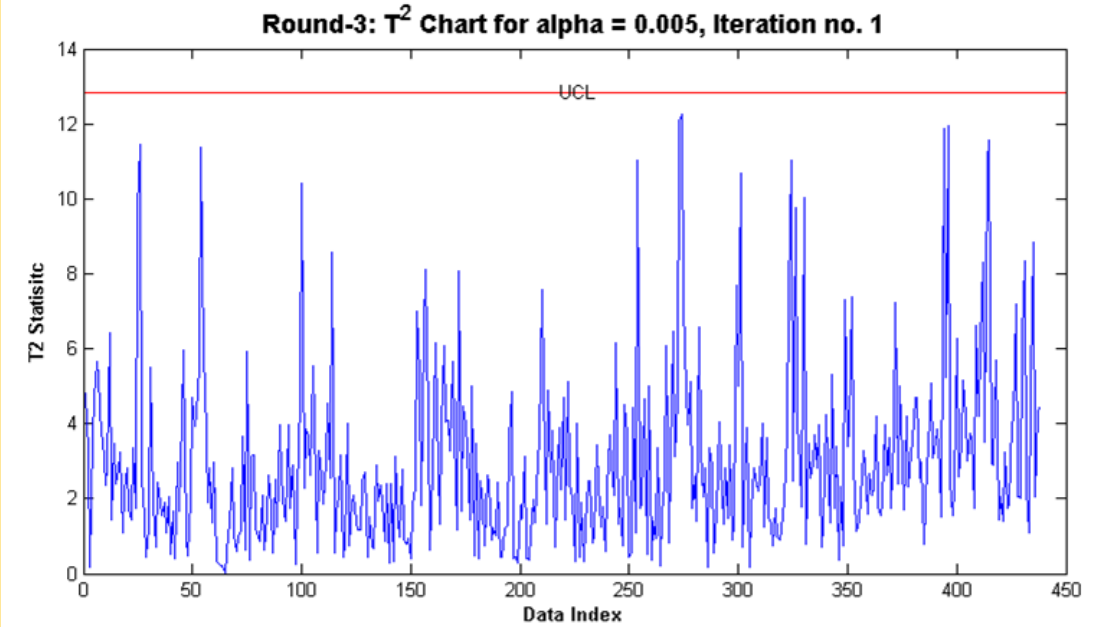


**Figure 18: $T^2$ Round 3 Iteration 1**

# Conclusion

- Isolating the OOC data from the IC data by doing multiple iterations of the $T^2$ and m-CUSUM control charts helped identify 6 more data points during the $2^{nd}$ round of iterations that were OOC
- Despite performing the PCA, the non-diagonal elements of the covariance matrix are non-zero, which implies the values are correlated and thus dependent
- This may be attributed to noise, which made it difficult to achieve zero covariance
- With the conclusion of Phase I analysis, the control charts established with the IC mean and covariance can be used to monitor future observations
- This project forced us to rely on our own judgment and to rationalize all methods used, similar to a possible industry scenario
- Having done the project we now have a better understanding of the relative advantages in utilizing one method over another