

# **Change Point Analysis and Applications**

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*24May2011*

# Topics

- Background
- Issue
- Questions
- Solutions
- Results

# Eli Lilly & Company

- Attention-deficit hyperactivity disorder
- Bipolar disorder
- Cancer
- Cardiology
- Depression
- Diabetes
- Diabetic Peripheral Neuropathic Pain

- Erectile dysfunction
- Fibromyalgia
- Growth disorders
- Osteoporosis
- Schizophrenia
- Sepsis
- Others

# Terms

- Manufacturing examples
- Drug Product (DP)
- Drug Substance (DS)
- Properties: many are measured
- Release values
- Criteria: change over time and variability

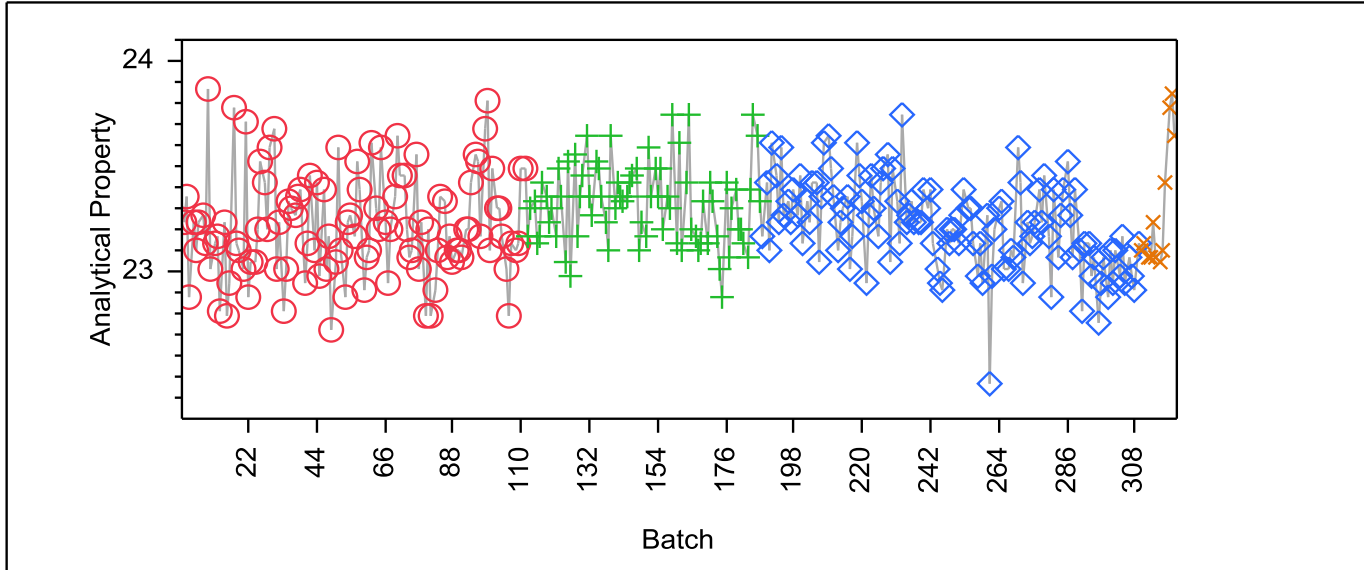
# Manufacturing

- ❑ DS batch → multiple DP batches
  - Potential for Autocorrelation
  - Not sequential
- ❑ DP batch – 500,000 dosage units
  - For properties, individual measurements per batch
- ❑ Measurements for incoming, inprocess, final product
- ❑ Process Analytical Tools (PAT)

# Business Case

- ❑ In manufacturing investigations, we often ask if and when a change in the process occurred. Given a level of confidence, we then use this information to be more efficient during the investigation.
- ❑ Control charts may indicate a change occurred, but it may not be clear exactly when.
- ❑ CUSUM charts help indicate a change occurred, but may be difficult to interpret

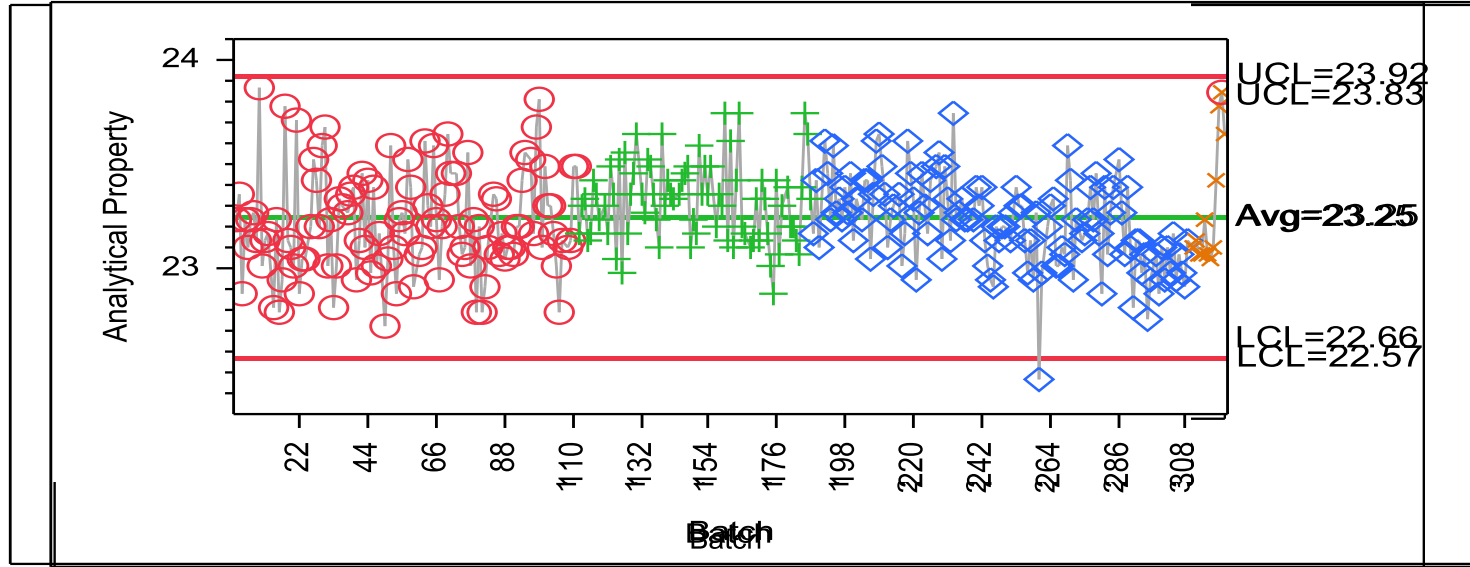
# Results1 – Run Chart



- Did a change occur?
- When did a change occur?
- Did more than one change occur?
- How confident are we that a change occurred?



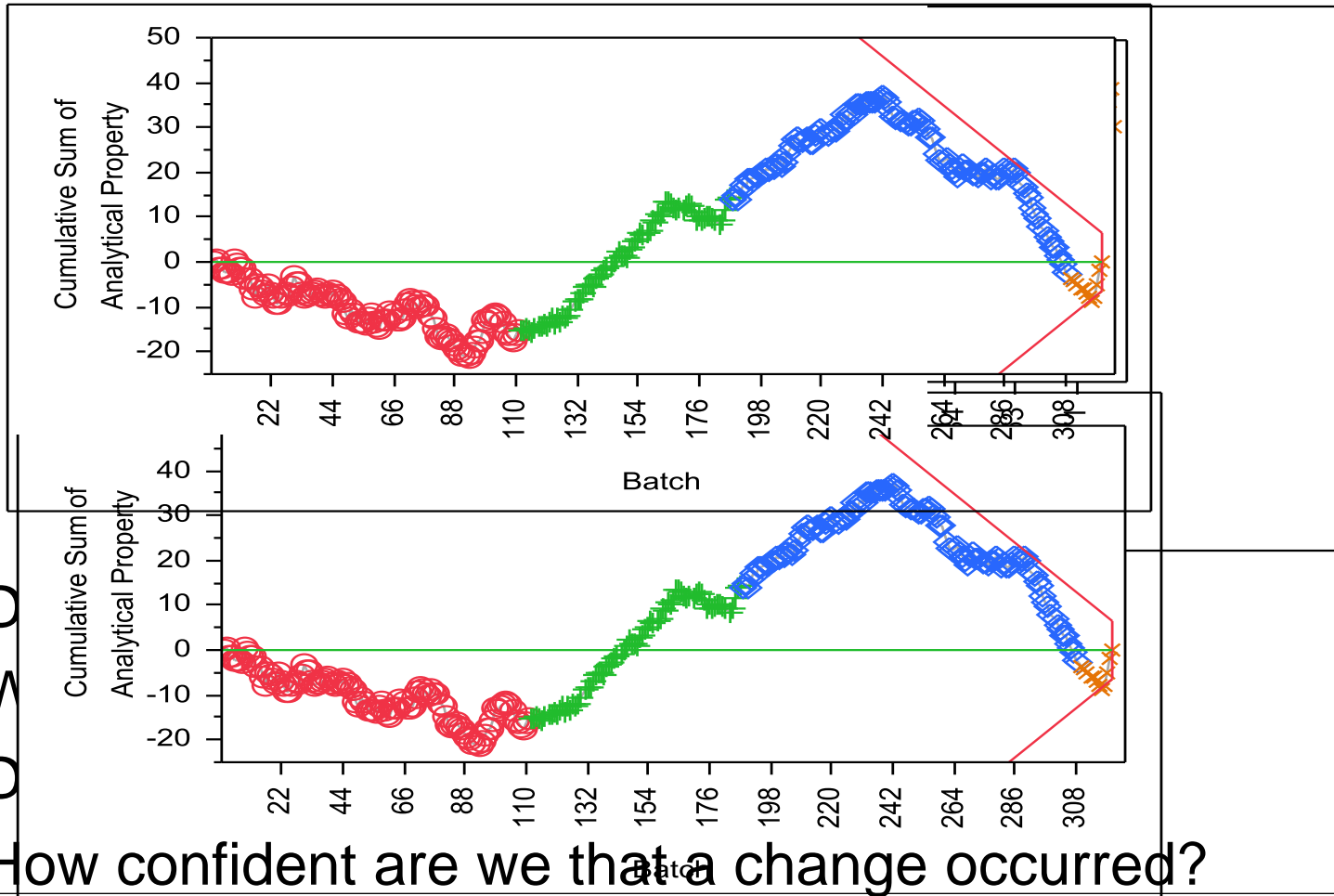
# Results1 - Control Chart



- Did a change occur?
- When did a change occur?
- Did more than one change occur?
- How confident are we that a change occurred?



# Results1 - CUSUM



# Change Point

- ❑ Based on paper by Dr. Wayne A. Taylor, 2000
  - Change-Point Analysis: A Powerful New Tool for Detecting Changes
- ❑ Designed to answer those questions
- ❑ Combines CUSUM and bootstrapping

# Steps

## □ Determine data set statistics

- Sort by date (mfg or test)
- Compute  $S_i = \text{lag } S_i + \text{result}_i - \text{average}$ ,  $i=1$  to  $n$
- Compute  $\text{Diff } S_0 = |\text{Max } S_i - \text{Min } S_i|$

## □ Determine confidence change has occurred

- Run 1000 simulations
- Randomize order (“shuffle” in JMP)
- Compute  $\text{Diff } S_k$ ,  $k=1$  to 1000
- Confidence =  $(\# \text{ Diff } S_k < \text{Diff } S_0) / 1000 \times 100\%$
- If confidence  $> 95\%$  → change has occurred

# Steps (con't)

- ❑ When did the change occur?
  - In original data set, find  $\max |S_i|$
  - Change point = location of  $\max |S_i| + 1$
- ❑ Divide dataset in two parts,  
(1) obs. 1 to  $\max |S_i|$  and (2)  $\max |S_i| + 1$  to  $n$
- ❑ Repeat procedure on each part to find additional changes (Level 2)
- ❑ Repeat

# JMP Script

```
// the identification of the Change Points
ChangeFlag1 = 0;
ChangeFlag11 = 0;
ChangeFlag12 = 0;
ChangeFlag21 = 0;
ChangeFlag22 = 0;
// Initial Change Level (used to update Change Column in the TC table) sets up to 1.
ChangeLevel = "1";

// TC is a new table for collecting data about the identified Change Points
// Lot Number, DR , DT , Confidence Level and Change Level (1, 2 or 3)
TC = New Table( "TC" );
TC << Add Rows( 1 );
TC << New Column( "DR" );
Column( "DR" ) << Format( "ddMonyyyy", 12, "ddMonyyyy" );
TC << Delete Column( "Column 1" );
TC << New Column( "DT" );
Column( "DT" ) << Format( "ddMonyyyy", 12, "ddMonyyyy" );
TC << New Column( "Lot", Character );
TC << New Column( "Confidence" );
Column( "Confidence" ) << Format( "Fixed Dec", 10, 1 );
TC << New Column( "Change", Character );

// TB is the reference to the table being analyzed. CUSUMs(S) and Abs(CUSUMs)(AbsS) are computed
TB = Open();
Current Data Table( TB );
New Column( "S", Formula( If( Row() == 1, :X - Col Mean( :X ), Lag( :S, 1 ) + (:X - Col Mean( :X )) ), Eval Formula );
New Column( "AbsS", Formula( Abs(:S), Eval Formula ));
TB << Run Formulas;

// Run Chart of the Initial data with identification of the detected Change Points.
// It is launched at the beginning of the script to save running time !!!
CC = Control Chart(Sample Size( :Lot ), KSigma( 3 ), Chart Col( :X, Run Chart( Show Center Line( 1 ), Test 1( 0 ), Test 2(
SendToReport(Dispatch({"Run Chart of X"}, "1", ScaleBox,
{Scale( Linear ), Format( "Best" ), Min( 0.5 ), Inc( 2 ), Minor Ticks( 0 ), Rotated Labels( 1 )}),
Dispatch( {"Run Chart of X"}, "Control Chart", FrameBox, Frame Size( 950, 350 ) ),
Dispatch( {"Run Chart of X"}, "Control Charts", FrameBox, Frame Size( 83, 350 ) ));

// Process Initial Table TB

// Create TS table for TB
```

Manipulate current data set

Add “shuffle” column

Add CUSUM column

Compute Diff; save to new data table

Current data set: Diff

New table data: Compute confidence

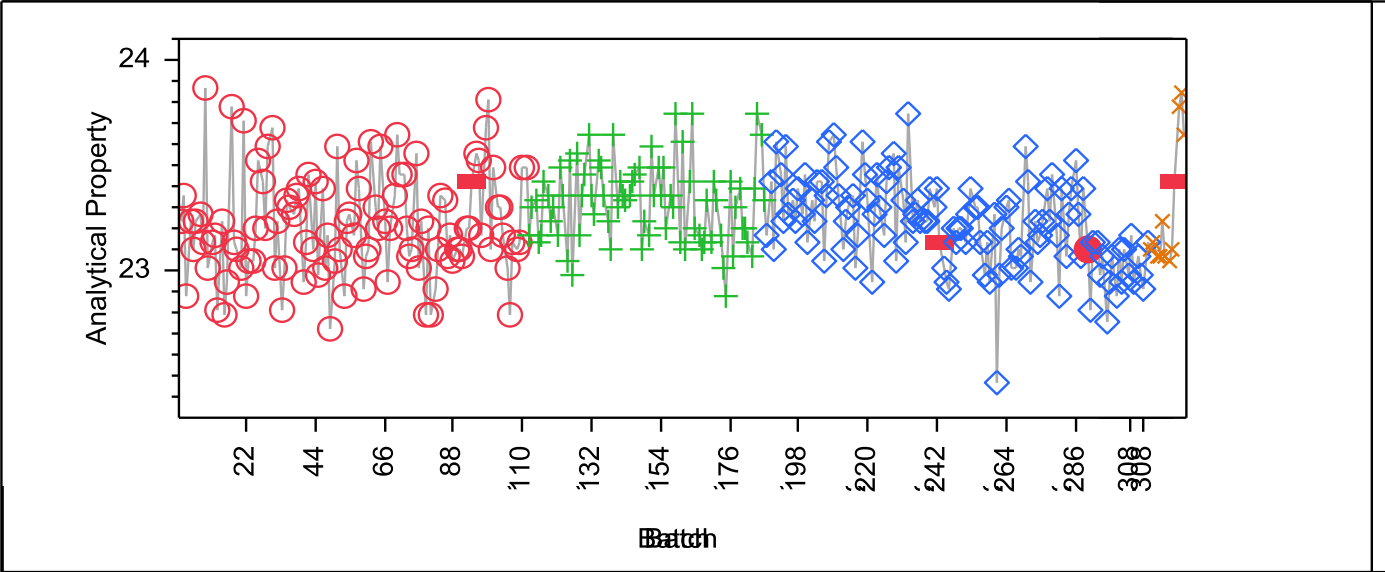
Current data set

Find change point

Split data set

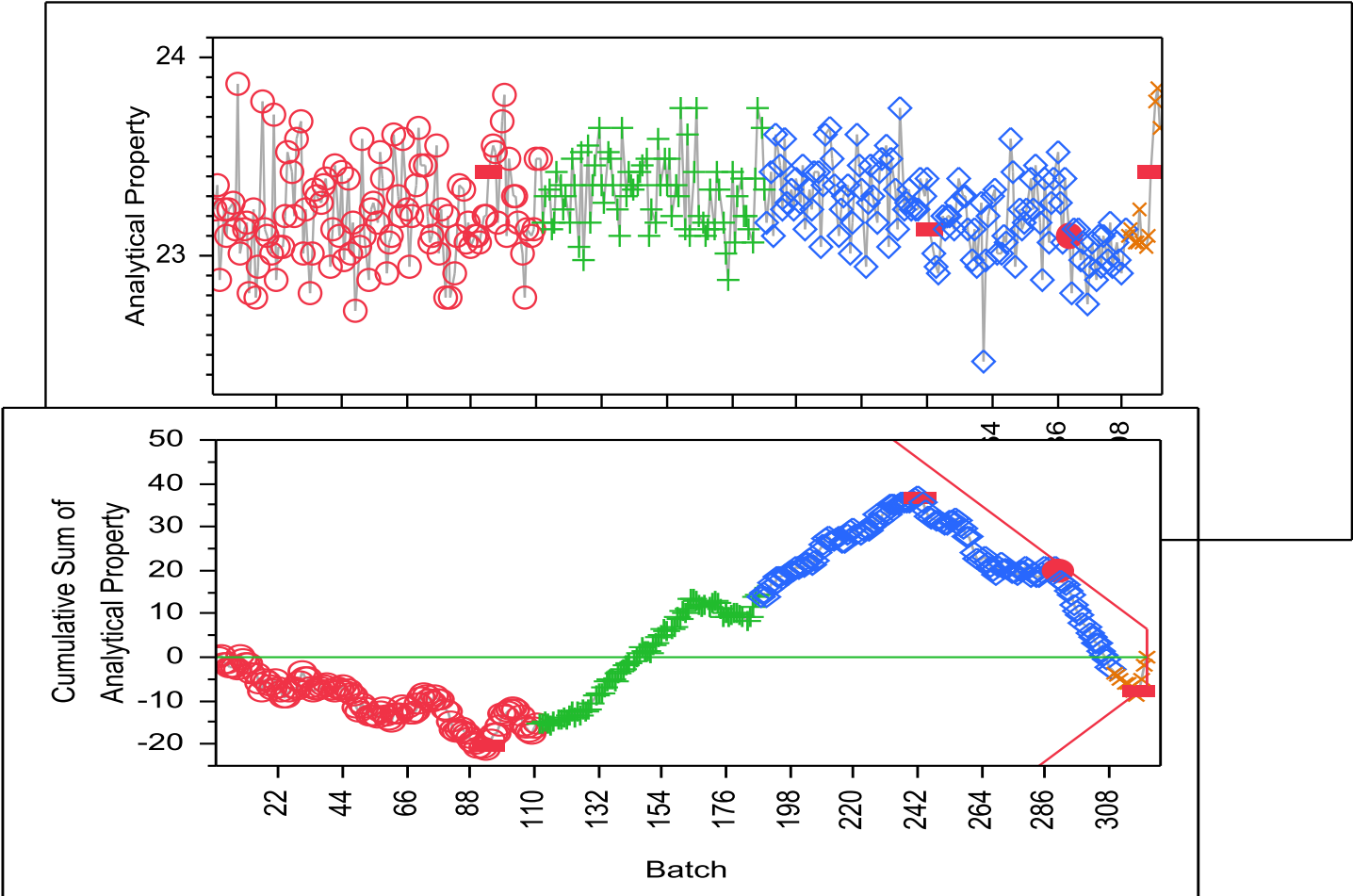
Repeat for each part of data set

# Results1 – Change Point



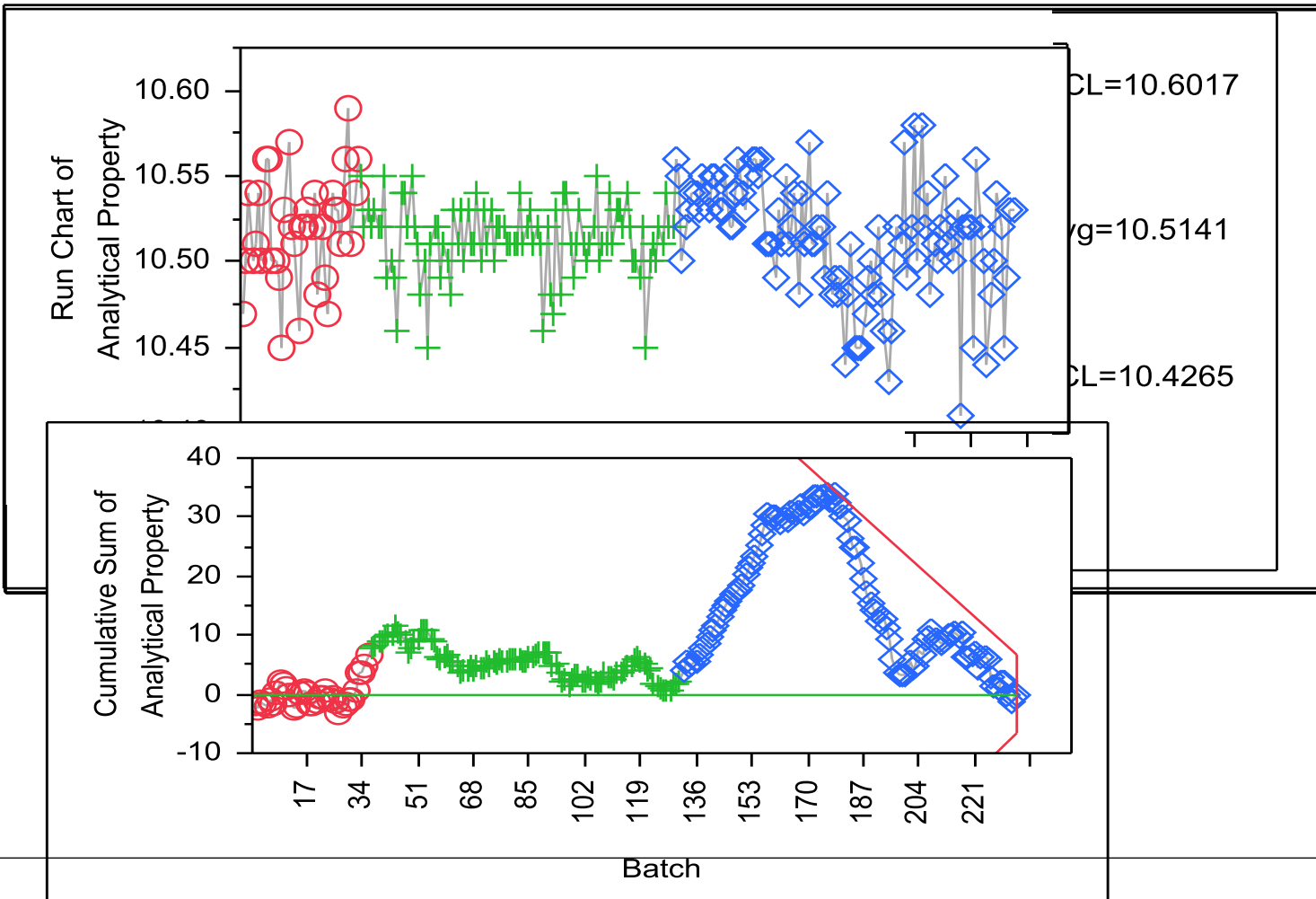
Batch	Confidence	Change
243	100.0	1
94	99.6	2
318	96.0	2
290	99.4	3

# Results1 - Comparison

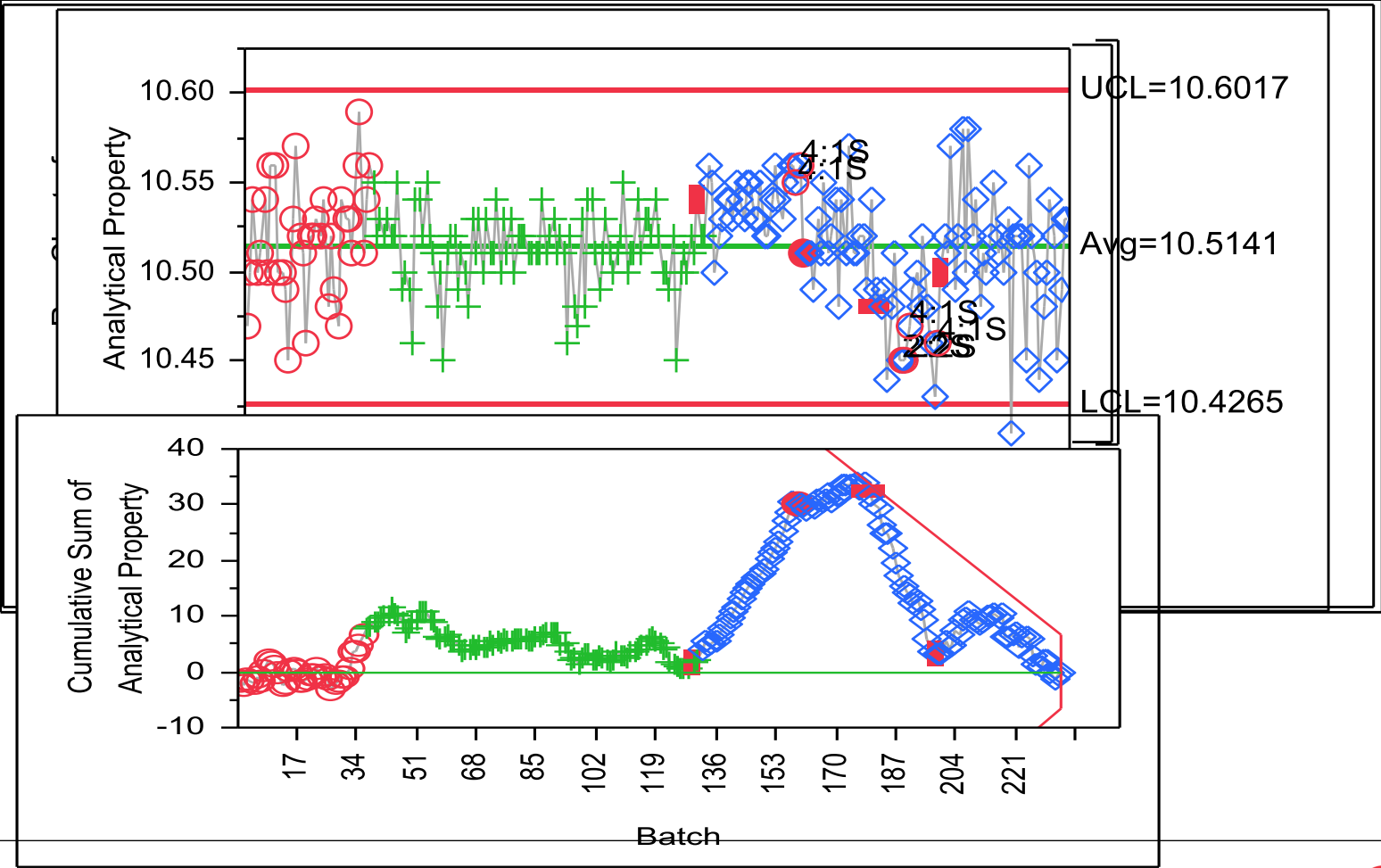




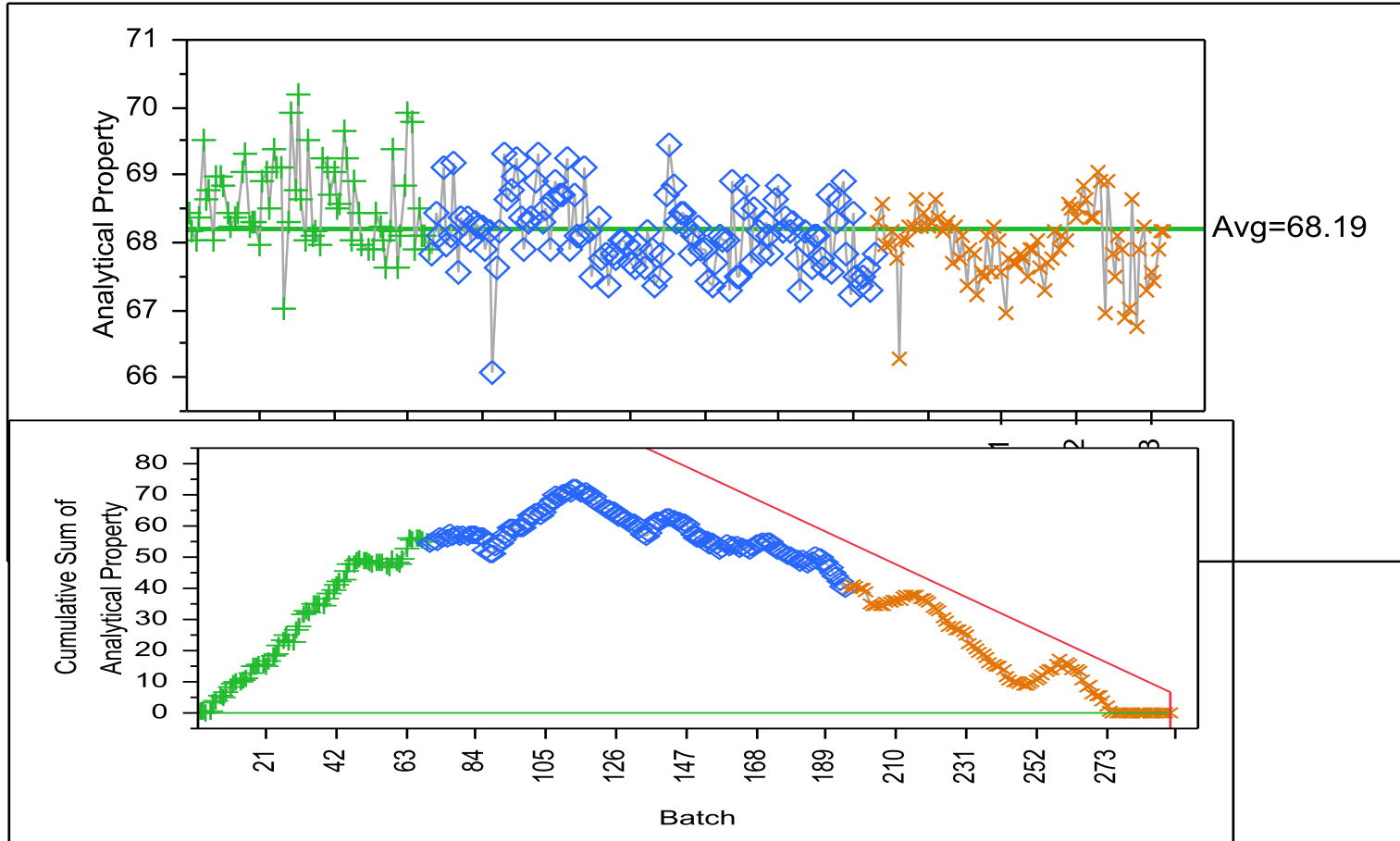
# Results2 – Raw Data



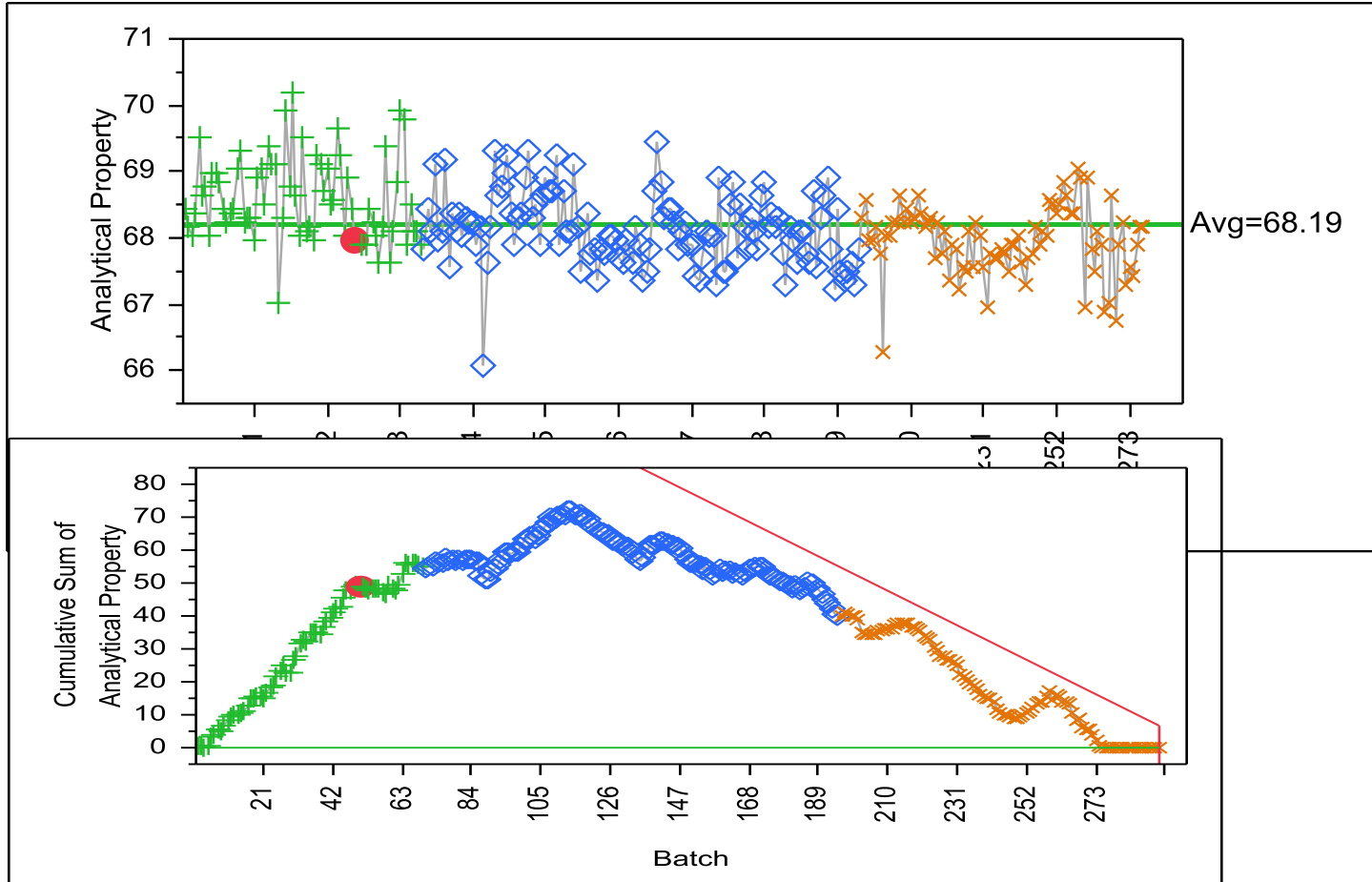
# Results2 – Change Point



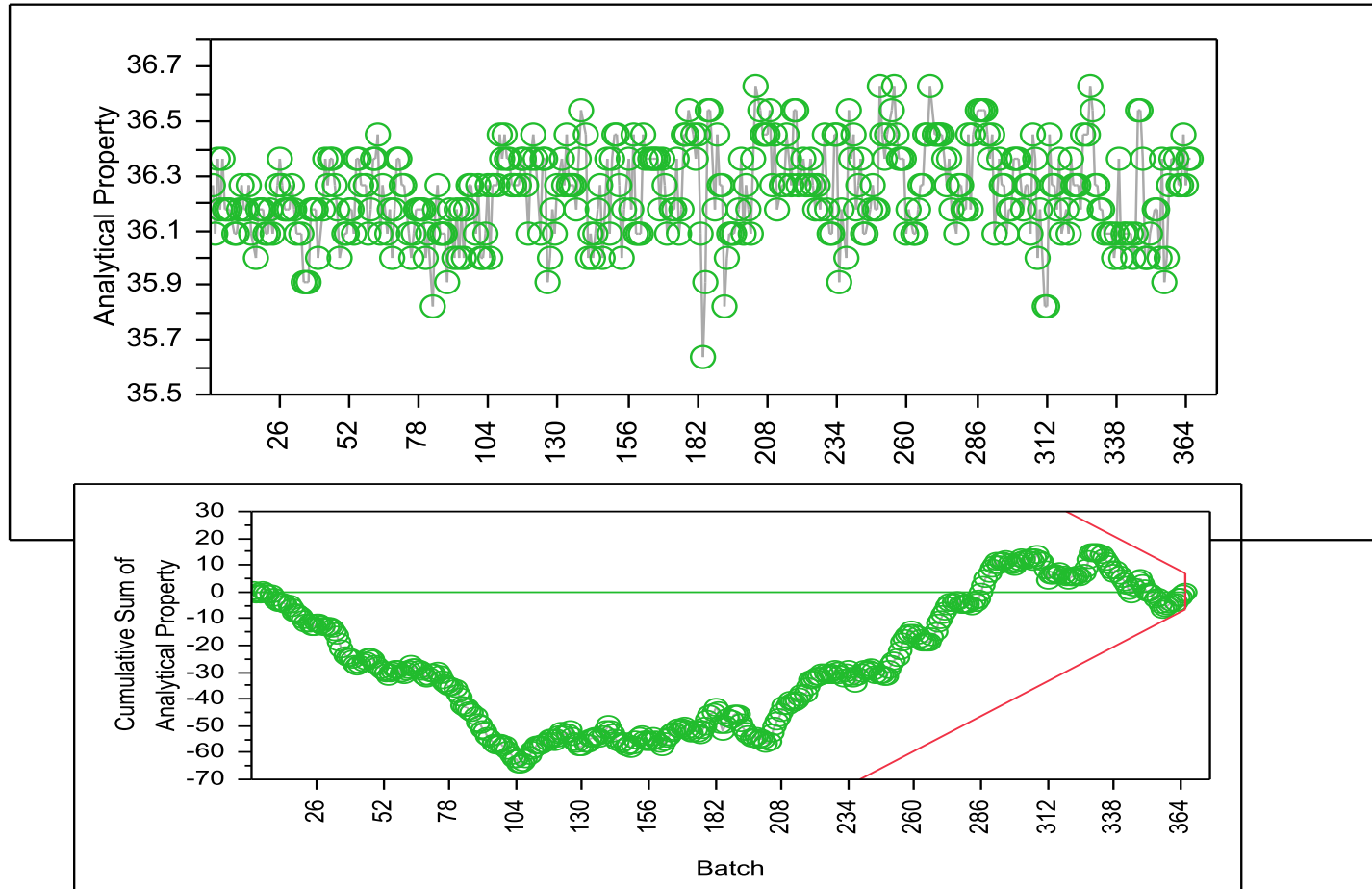
# Results3 – DP Raw Data



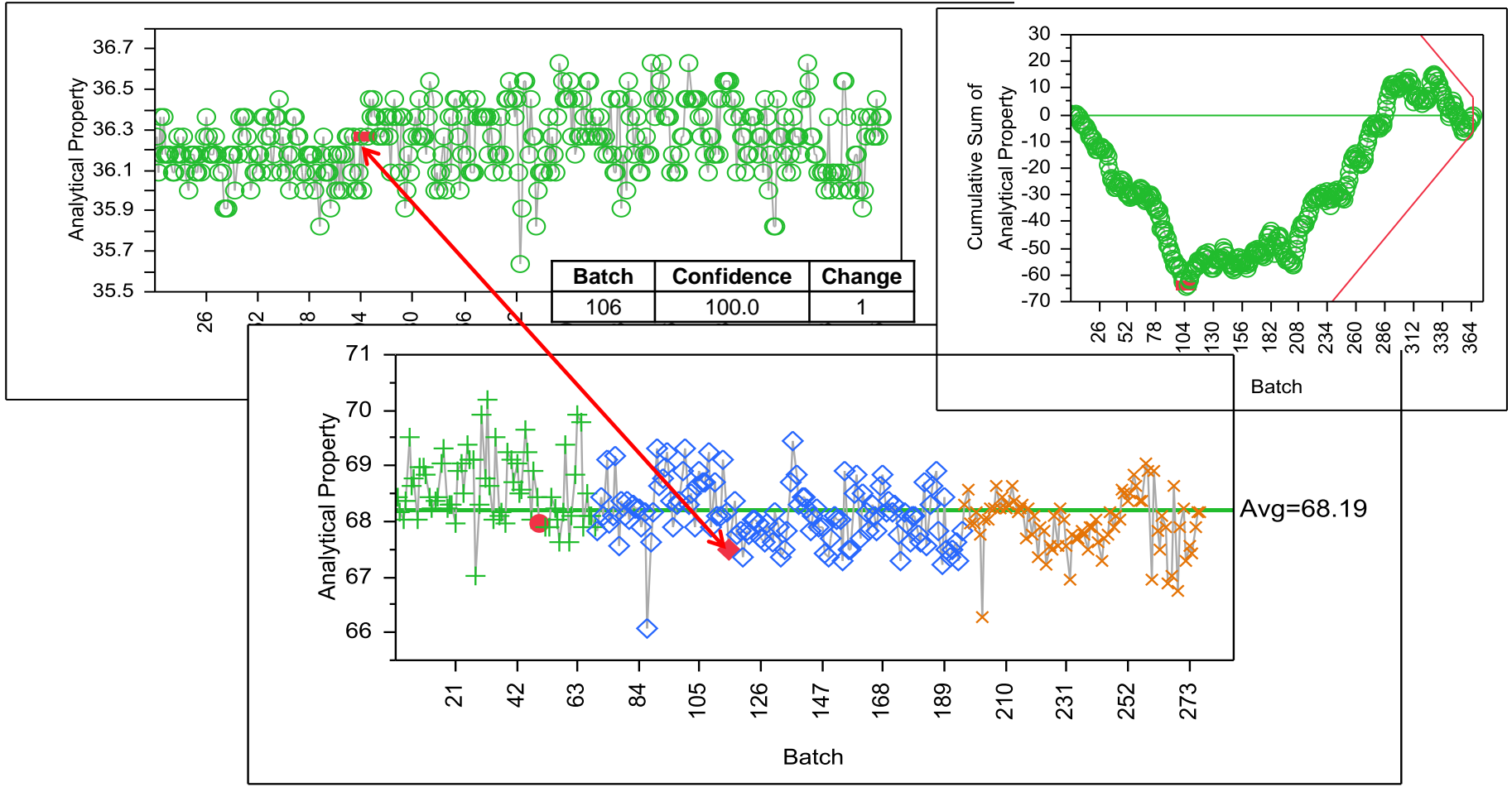
# Results3 – DP Change Point



# Results3 – DS Raw Data



# Results3 – DS DP Change Point





# Advantages

- ❑ Useful in verifying improvements, problem solving and in trend analysis
- ❑ Better characterizes changes including detection of multiple changes, providing associated confidence levels, and providing confidence intervals for the times of the changes.
- ❑ Flexible: no specific distribution is assumed: non-normal, ill-behaved (particle counts, complaint data), ranks (robust to outliers)
- ❑ Powerful at detecting subtle, sustained changes as with CUSUM vs control charts
- ❑ Simpler to use and easier to interpret than CUSUM charts.
- ❑ Controls change-wise error rate so reduces the number of false detections when dealing with large data sets.



# Disadvantages

- ❑ Not monitoring tool
- ❑ Not efficient at detecting isolated abnormal points
- ❑ Requires some independence, at least no strong autocorrelations
- ❑ Bootstrapping may result in different results
- ❑ Not replacement for control charts, use in conjunction

# Considerations

- Outliers
  - Use ranks instead
- Other estimators
  - MSE
- Bootstrap confidence intervals on when change occurred

# Speakers

## Hesham M. Fahmy

In this presentation, we review statistical procedures used to detect change points in manufacturing processes. Change point detection procedures provide process engineers with useful information in their search for special causes and could significantly reduce the cost and time of the identification process. We analyze and compare the performance of cumulative sum (CUSUM) and exponentially weighted moving average (EWMA) change point estimation procedures. A case study is presented to demonstrate the effectiveness of CUSUM and EWMA procedures to identify change points.

## Mark Holland

An extensive study has been conducted with the purpose of detecting a shift in location when univariate observations are collected. Many techniques have also been proposed to detect a shift in location vector when each observation consists of multiple measurements. These procedures require the user to make assumptions about the distribution of the process readings, to assume that process parameters are known, or to collect a large training sample before monitoring the ongoing process for a change in distribution. We propose a nonparametric procedure for multivariate statistical process control without making those assumptions or collecting a large training sample.