

Mudslide Mitigation

Chunxi Wang; Yi Yang; Zhaoxing Zhang; Zheng Tang

Group5

Syracuse University

Department of Mechanical and Aerospace Engineering

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Problem Statement

- Extremely rapid surging flow
- Significant proportion of clay
- Cause serious damage
- Happens in a short time



Problem Statement

mudslide usually occurs quickly and with great uncertainty, leaving a short response time, so it is easy to cause problems in the disaster management process.

- Mudslide is difficult to predict and prevent
- Extreme weather conditions hampered escape and rescue efforts
- As a result of the lack of early training, when the disaster occurred personnel confusion
- Emergency supplies are in short supply
- Reconstruction after the disaster is difficult

Previous case

Tibes, Ponce, Puerto Rico

Caused by heavy rainfall from Tropical Storm Isabel in 1985. The mudflow destroyed more than 100 homes and claimed an estimated 300 lives.



Previous case

Zhouqu, Gansu, China

Occurred at 12 midnight on 8 August 2010. It was caused by heavy rainfall and flooding in Gansu Province.

This mudslides killed more than 1,471 people as of 21 August 2010, while 1,243 others have been rescued and 294 remain missing.

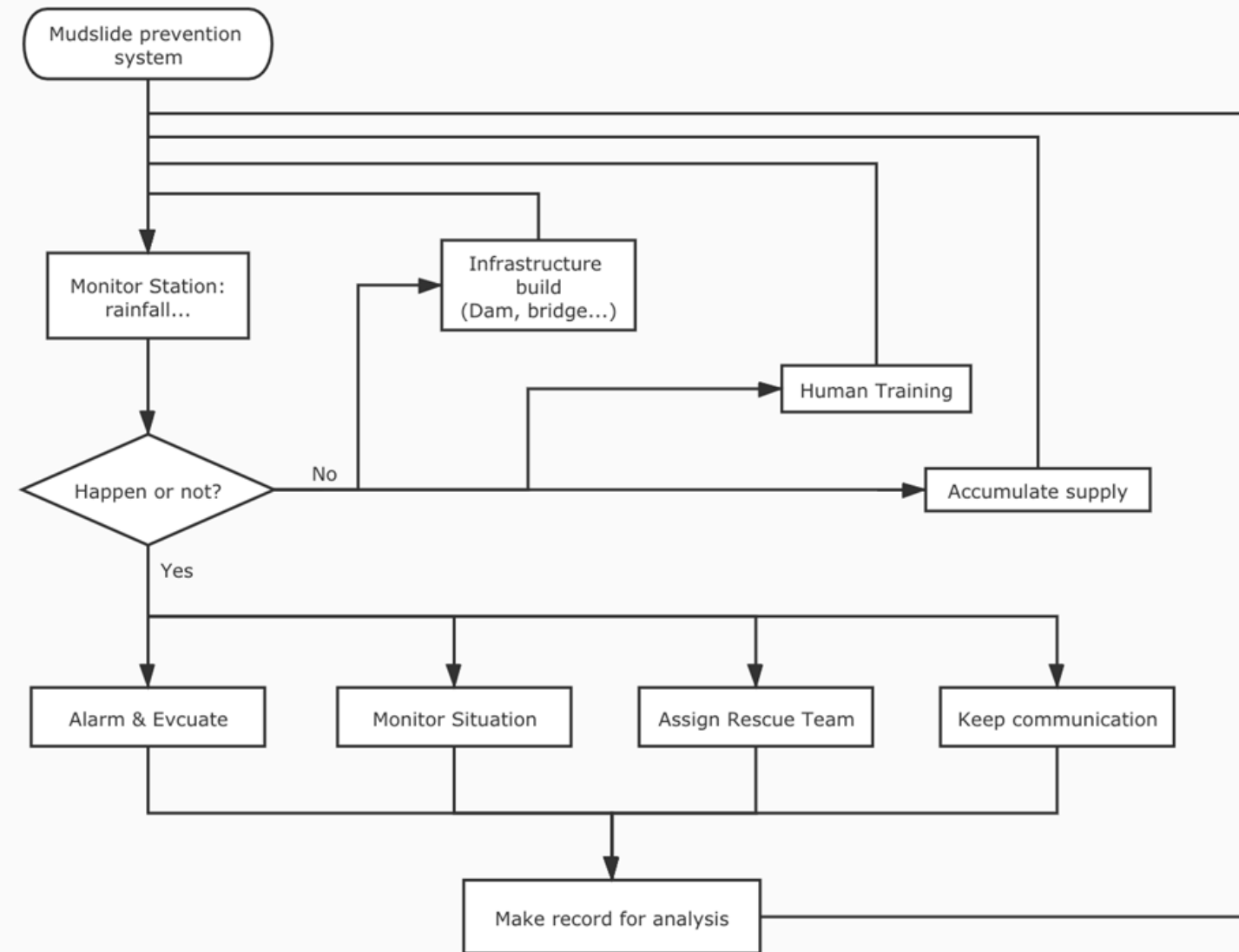
Around 1,800,000 cubic metres (64,000,000 cu ft) of mud and rocks swept through the town



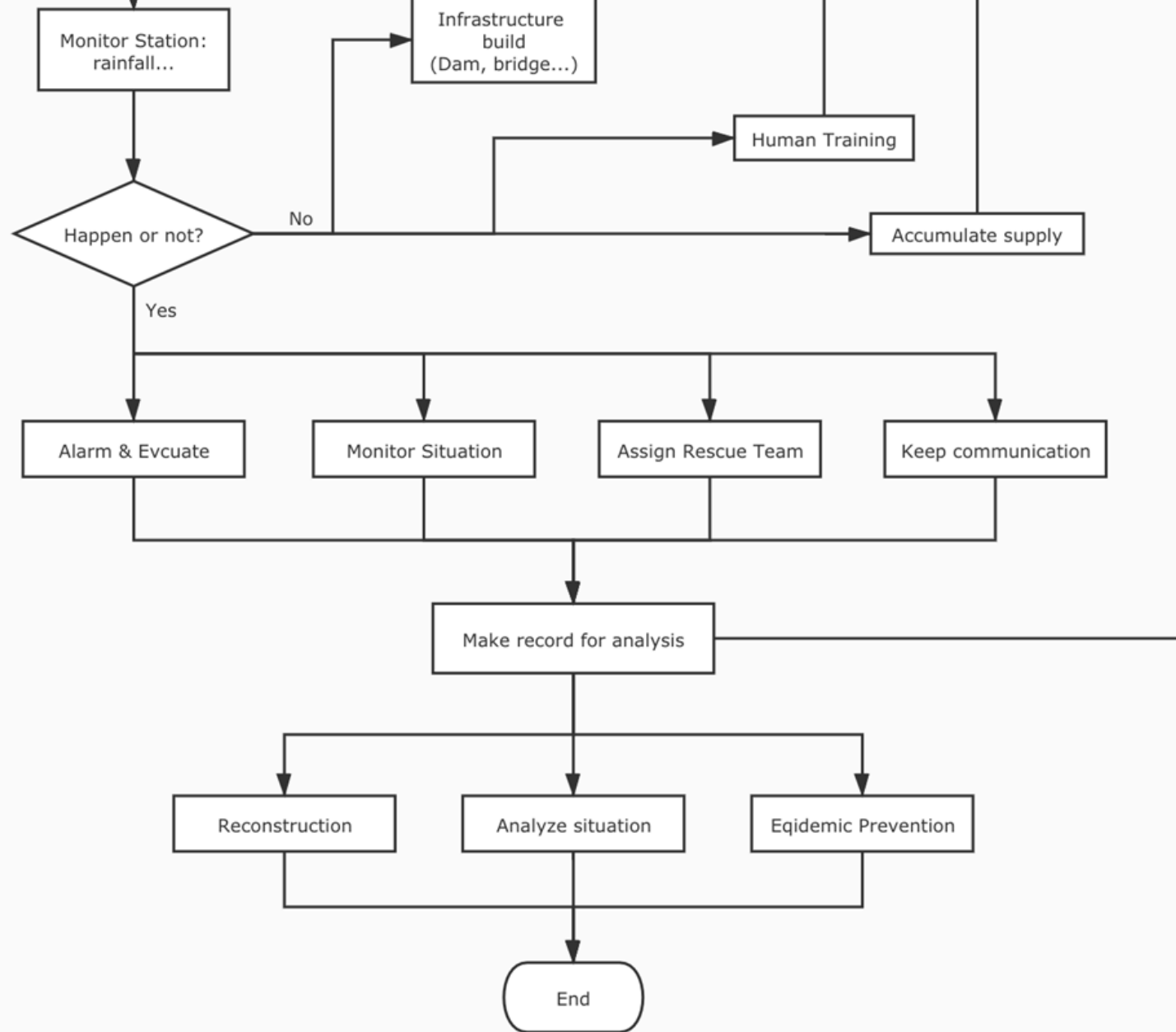
Brainstorming Session

Before	During	After
Detection	Alarm	Reconstruction
Preparation	Monitoring	Analyze
Training	Rescue	Epidemic prevention
	Communication	

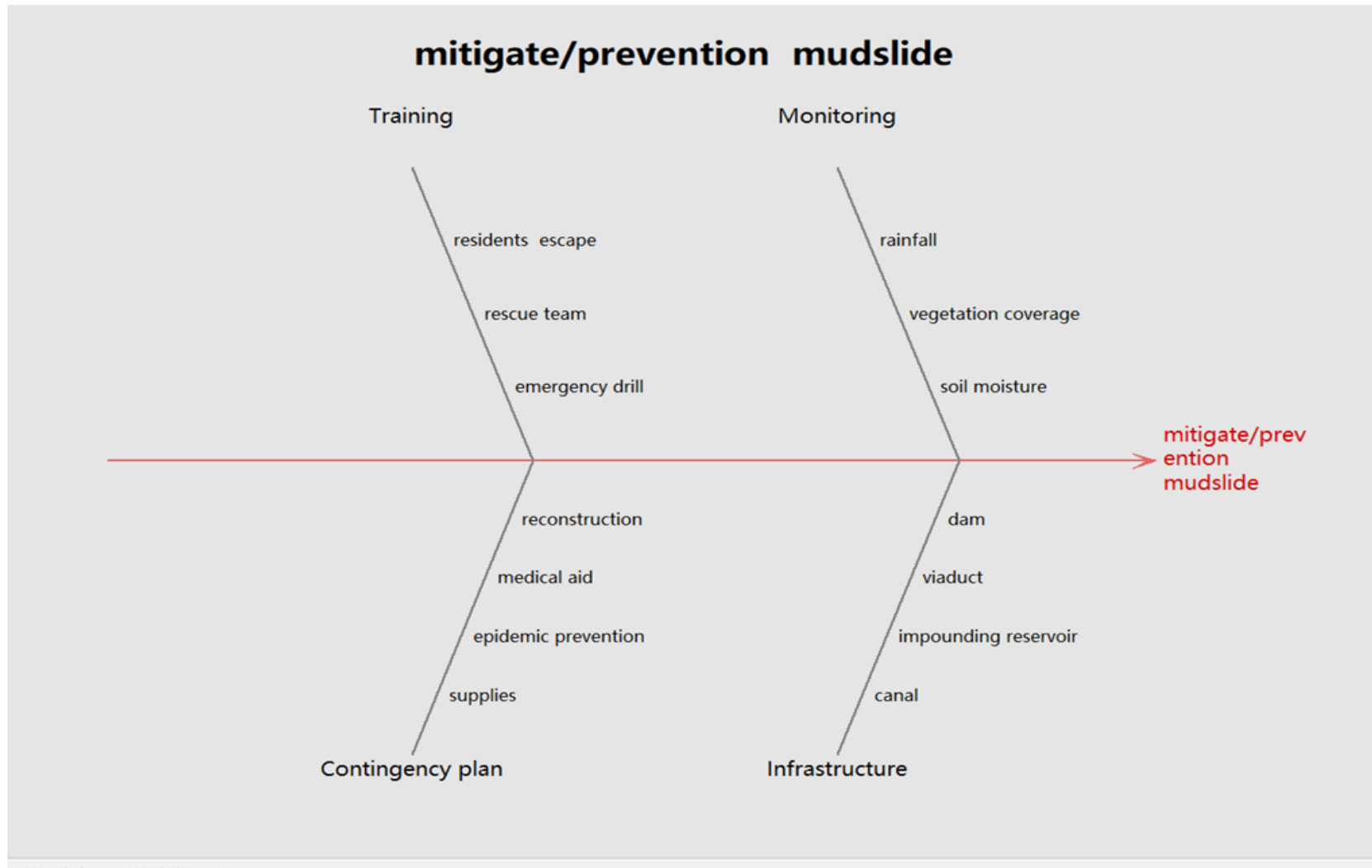
Flow chart



Flow chart



Fishbone chart



COPQ

Cost of Poor Quality				
Process	Internal Failure Costs	External Failure Costs	Appraisal Costs	Prevention Costs
Monitoring Mud Slide	Failure of equipment	Breach of duty	Rules	Inspect and equipment maintaining
Local hospital prepared	Failure of equipment/lack of doctors and room	Destroyed/electricity ruin	Monitoring	Improve medical service and electricity back up
Infrastructure	Lack of Dam, bridge, canal, sludge storage	Few workers	Monitoring	Infrastructure building
Evacuation people/ Transfer injured	Lack of vehicle/ambulance	Roads be destroyed	Method(human training) and spare place, amount of the vehicle	Train people how to get to safe place and evacuation method
Keeping Communication	No enough signal base stations	Signal interference	Signal base stations maintaining	Pay more on advanced and stable equipment
Epidemic Prevention	Too late to transfer dead people and animals	No enough vehicle	Laws	Emergency epidemic prevention plan(training firefighters)

6 SIGMA

DMAIC

DMAIC is a five-step method for improving existing process problems with unknown causes.



Define

Problem Statement

- Dams are unable to stop or slow down the mudslide;
- Capability of canal is not enough for mudslide to flow;
- Insufficient food and medical supplies;
- People cannot evacuate in time.

Measure

Data Collection

	Where will data be collected	Who will provide the data	How often
Amount of Rainfall		Weather Department	Every time it rains
Moisture of soil on mountains	On the mountain	Geology Department	Before and during rain
Amount of mud flushed down the mountain	In the canal	Reconstruction Team	After the mudslide
Local population		Local government	Every year
Amount of supplies (food, medicine, etc.) used during a single mudslide		Local government	After the mudslide

Gage R&R Analysis

Gage R&R--ANOVA Method

Gage R&R Study - ANOVA Method

Two-Way ANOVA Table With Interaction

Source	DF	SS	MS	F	P
Part	9	88.3619	9.81799	492.291	0.000
Operator	2	3.1673	1.58363	79.406	0.000
Part * Operator	18	0.3590	0.01994	0.434	0.974
Repeatability	60	2.7589	0.04598		
Total	89	94.6471			

α to remove interaction term = 0.05

Two-Way ANOVA Table Without Interaction

Source	DF	SS	MS	F	P
Part	9	88.3619	9.81799	245.614	0.000
Operator	2	3.1673	1.58363	39.617	0.000
Repeatability	78	3.1179	0.03997		
Total	89	94.6471			

Gage R&R

Variance Components

Source	VarComp	%Contribution (of VarComp)
Total Gage R&R	0.09143	7.76
Repeatability	0.03997	3.39
Reproducibility	0.05146	4.37
Operator	0.05146	4.37
Part-To-Part	1.08645	92.24
Total Variation	1.17788	100.00

Gage Evaluation

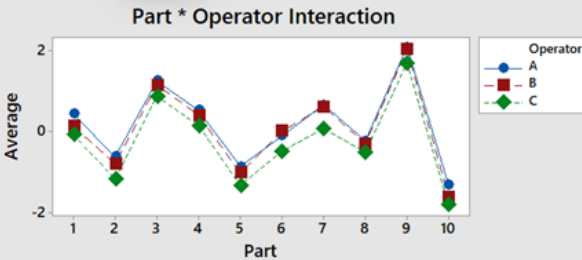
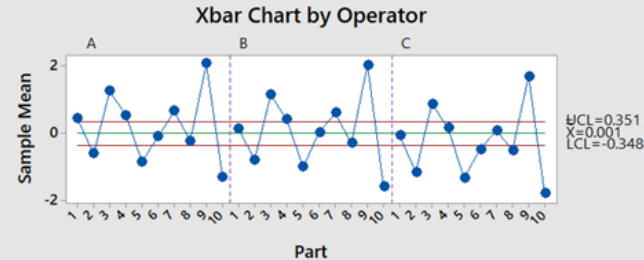
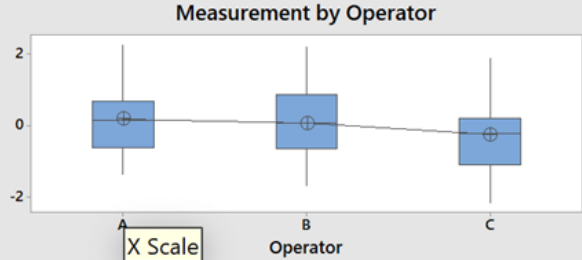
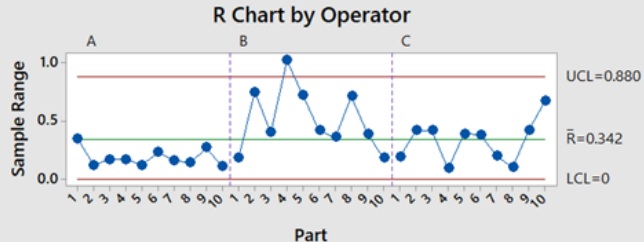
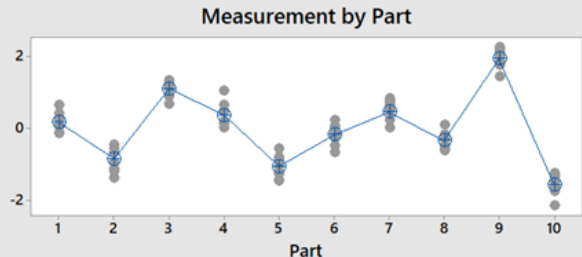
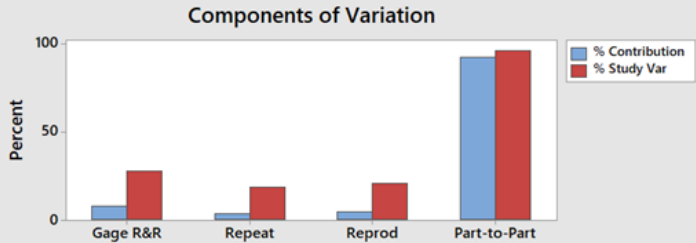
Source	StdDev (SD)	Study Var (6 × SD)	%Study Var (%SV)
Total Gage R&R	0.30237	1.81423	27.86
Repeatability	0.19993	1.19960	18.42
Reproducibility	0.22684	1.36103	20.90
Operator	0.22684	1.36103	20.90
Part-To-Part	1.04233	6.25396	96.04
Total Variation	1.08530	6.51180	100.00

Gage R&R Analysis

Gage R&R (ANOVA) Report for Measurement

Gage name:
Date of study:

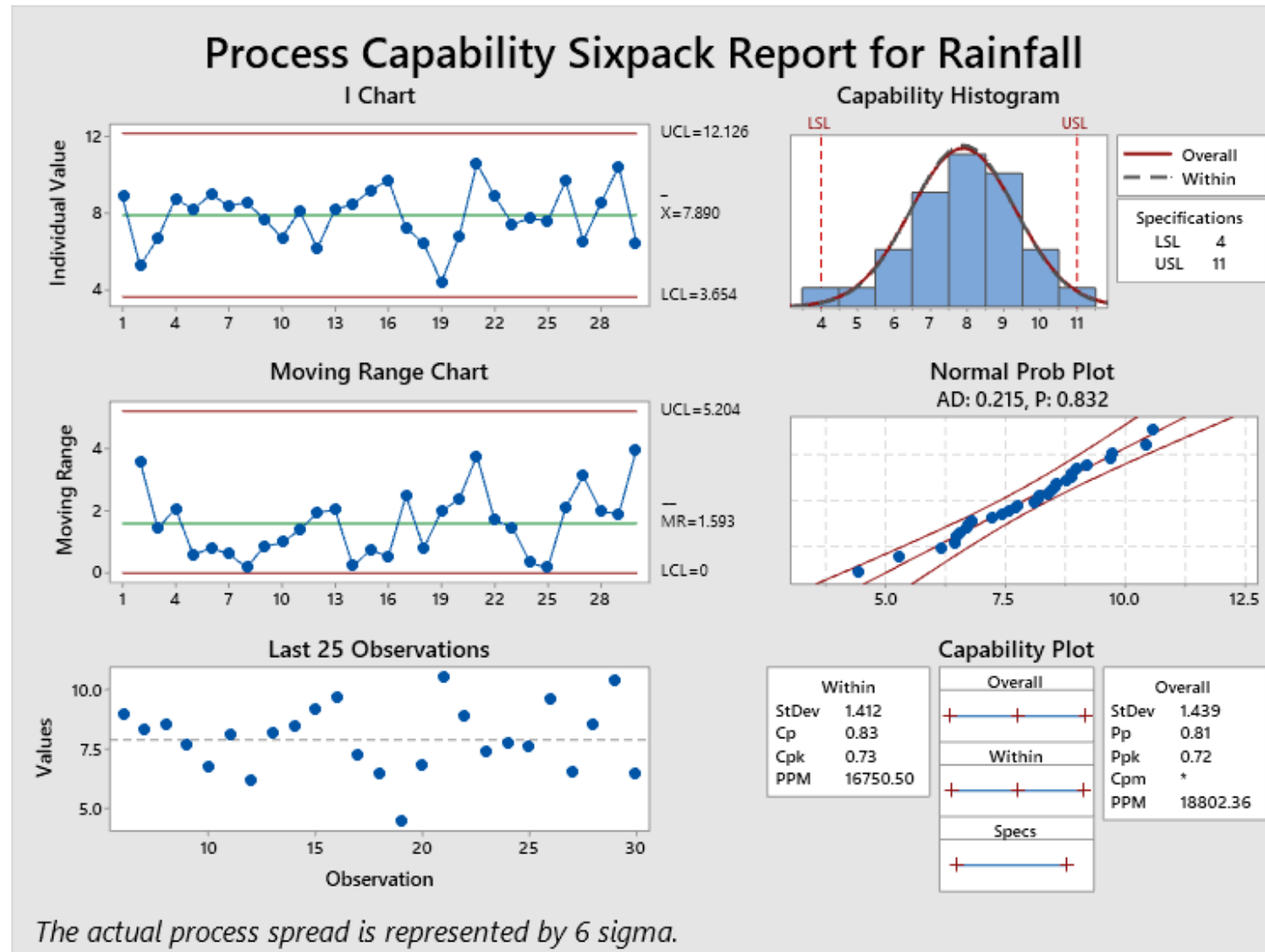
Reported by:
Tolerance:
Misc:



Analyse

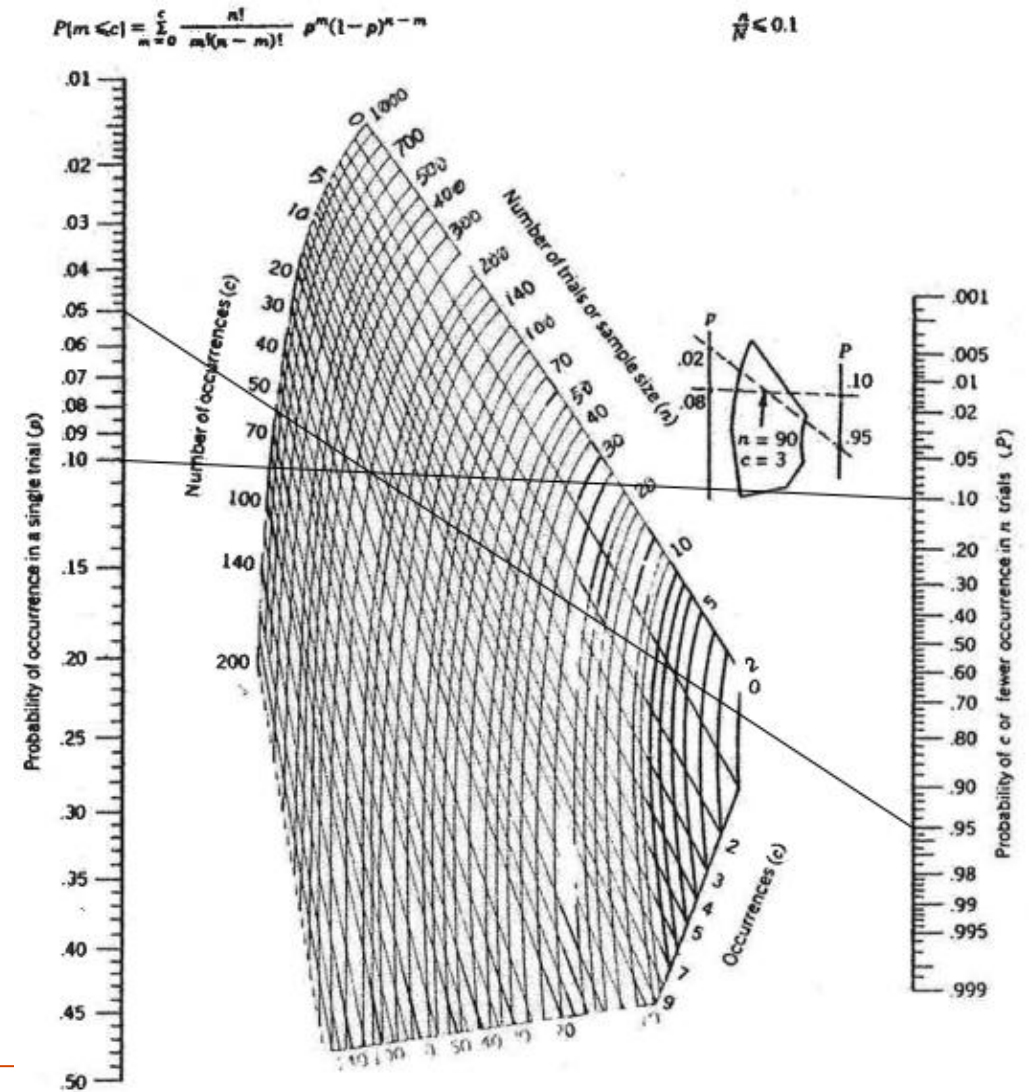
Rainfall (inches)		
8.86707	8.114762	10.57913
5.265968	6.174741	8.882498
6.696338	8.212439	7.4181
8.773968	8.44493	7.74245
8.183697	9.180771	7.575326
8.984158	9.717632	9.678482
8.367323	7.22412	6.550373
8.551716	6.426168	8.538369
7.709114	4.423975	10.43627
6.727064	6.797707	6.462808

Analyse



Acceptance Sampling

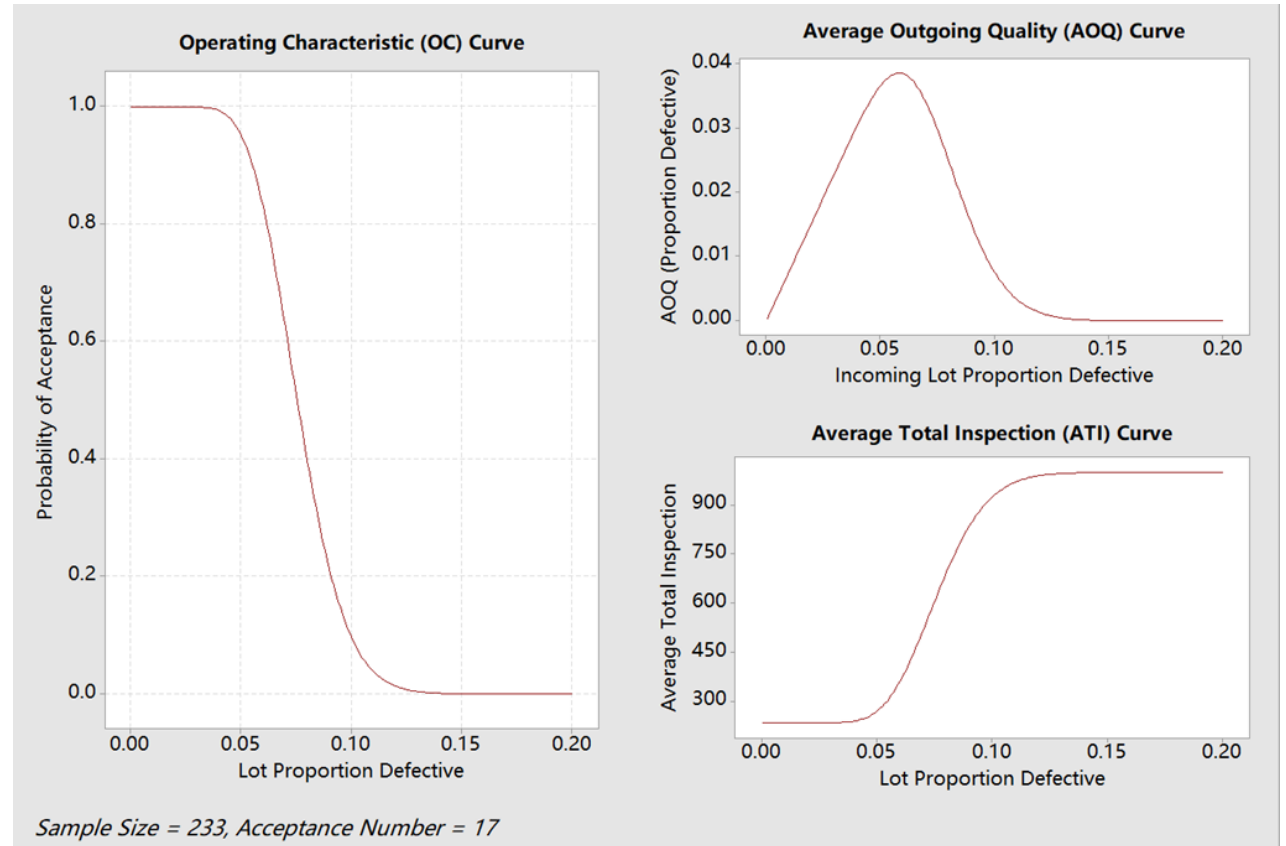
AQL=0.05
 LTPD=0.1
 $\alpha=0.05$
 $\beta=0.1$



Operating Characteristic Curve(OC Curve)

An OC curve visualizes the probability for a sampling plan, showing the probability of accepting a lot given the percent defectiveness.

This probability is calculated using a binomial distribution.



Operating Characteristic Curve(OC Curve)

Acceptance Sampling by Attributes

Measurement type: Go/no go
 Lot quality in proportion defective
 Lot size: 1000
 Use binomial distribution to calculate probability of acceptance

Method

Acceptable Quality Level (AQL) 0.05
 Producer's Risk (α) 0.05
 Rejectable Quality Level (RQL or LTPD) 0.1
 Consumer's Risk (β) 0.1

Generated Plan(s)

Sample Size 233
 Acceptance Number 17

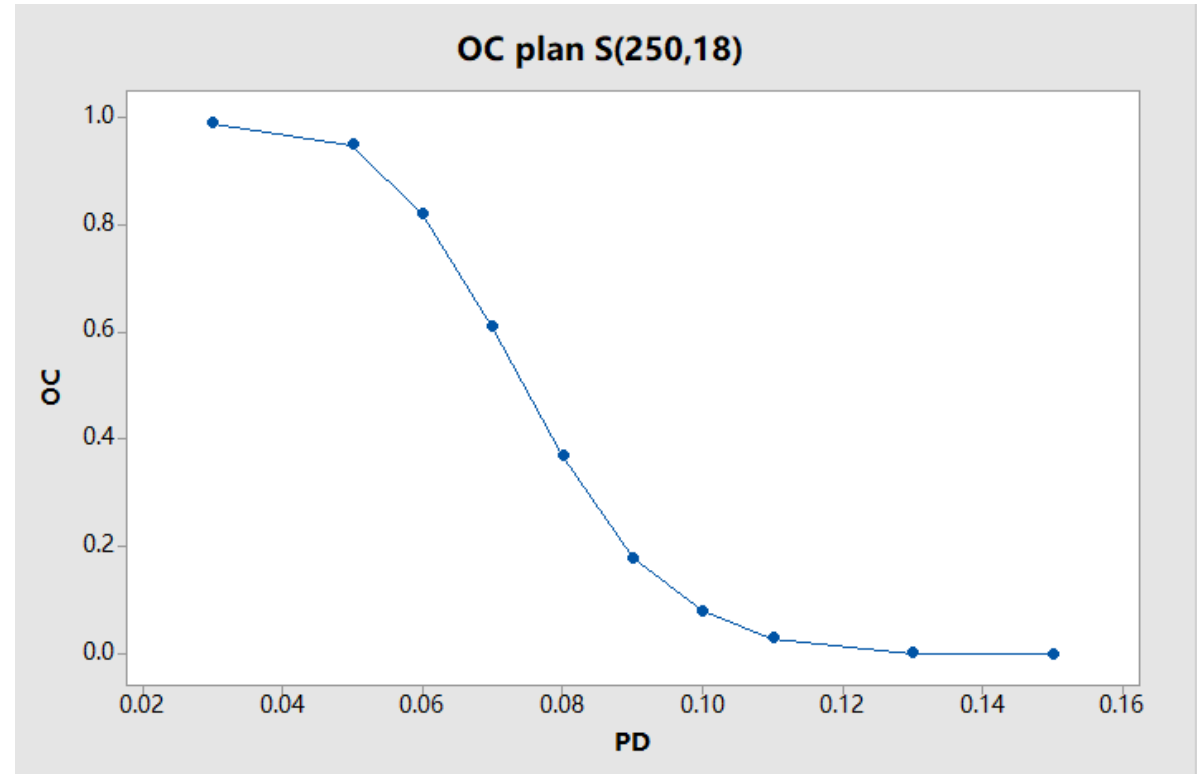
Accept lot if defective items in 233 sampled ≤ 17 ; Otherwise reject.

Proportion Defective	Probability Accepting	Probability Rejecting	AOQ	ATI
0.05	0.954	0.046	0.03658	268.3
0.10	0.099	0.901	0.00758	924.2

Average Outgoing Quality Limit(s) (AOQL)

AOQL	At Proportion Defective
0.03855	0.05815

Graphs - Acceptance Sampling by Attributes



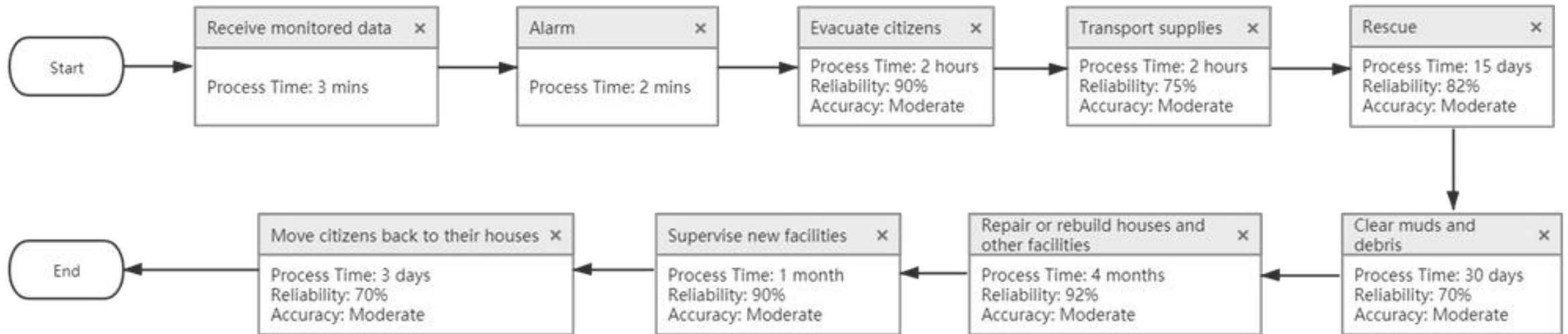
Improve

Provide solutions to the problem

- Improvement Strategy
- Design of Experiments (DOE)
- Value Stream Map (VSM)
- Reliability Analysis
- List of remedies selected

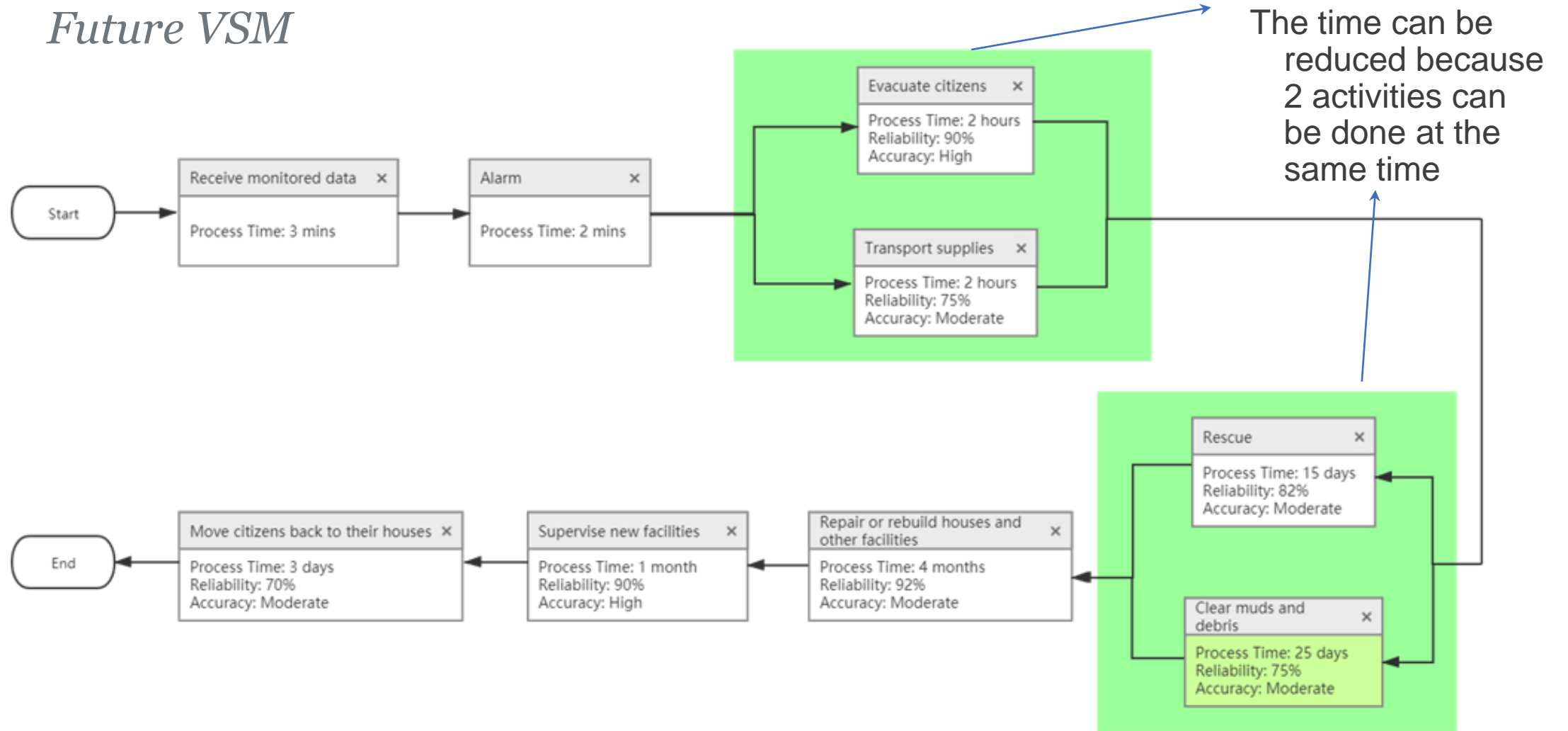
VSM

Current VSM



VSM

Future VSM



Design of Experimental

For this experiment we are conducting a full factorial design. That is we have 3 factors (Time, Temp and Catalyst) each of which have two levels and 2 replicates.

The concept of coding is used to both differentiate between high and low values and to determine later values. Coding is simple taking either the High or Low value subtracting the midpoint, divided by the range and then multiplying times two. This normalization is done to ensure a standardized combination of factors.

Factor	Low level	High level
time	20	50
temp	150	200
catalyst	A	B

	Low level 20-34	High level 35-50
time	-1	1
	Low level 150-174	High level 174-200
temp	-1	1
	Low level A	High level B
catalyst	-1	1

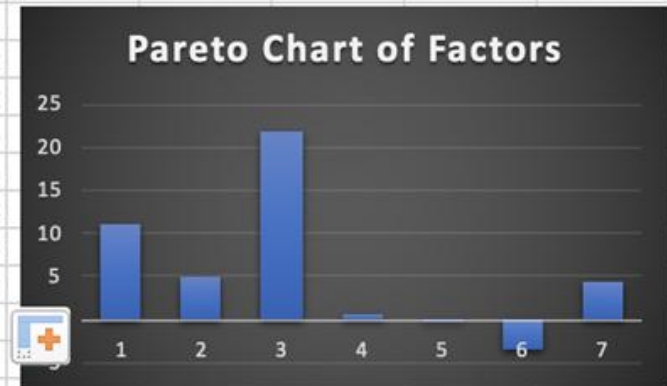
DOE DATA AND CODING

Factorial Experiments 2 ³ (DOE)									
Runs	A	B	C	AB	AC	BC	ABC	Y1	Y2
1	-1	-1	-1	1	1	1	-1	0.51817	-0.49313
2	1	-1	-1	-1	-1	1	1	6.988546	8.020517
3	-1	1	-1	-1	1	-1	1	5.674872	6.574334
4	1	1	-1	1	-1	-1	-1	6.500394	13.32876
5	-1	-1	1	1	-1	-1	1	15.67002	14.22132
6	1	-1	1	-1	1	-1	-1	15.96954	19.63775
7	-1	1	1	-1	-1	1	-1	16.34648	9.913479
8	1	1	1	1	1	1	1	20.72943	21.31062

Excel DOE

Excel can be used to create the framework for this experiment. Below we have created an experiment design where we have our 3 factors, our 2 levels and our 16 runs. Note that Y1 and Y2 represent the responses for the runs.

Run	Factorial Experiments 2 ³ (DOE-ASQ)							Run Results			
	A	B	C	AB	AC	BC	ABC	Y1	Y2	Avg.	Var.
1	-1	-1	-1	1	1	1	-1	0.51817	-0.49313	0.01252	0.511365
2	1	-1	-1	-1	-1	-1	1	6.988546	8.020517	7.504531	0.532482
3	-1	1	-1	-1	1	1	-1	5.674872	6.574334	6.124603	0.404516
4	1	1	-1	1	-1	-1	-1	6.500394	13.32876	9.914578	23.3133
5	-1	-1	1	1	1	-1	1	15.67002	14.22132	14.94567	1.049371
6	1	-1	1	-1	1	1	-1	15.96954	19.63775	17.80364	6.727862
7	-1	1	1	-1	-1	-1	1	16.34648	9.913479	13.12998	20.69173
8	1	1	1	1	1	1	1	20.72943	21.31062	21.02003	0.168893
TotSum								88.39745	92.51365	90.45555	53.39952
SumY+	56.24278	50.18919	66.89932	45.89279	44.9608	41.66706	49.59483				
SumY-	34.21277	40.26636	23.55623	44.56276	45.49475	48.78849	40.86072				
AvgY+	28.12139	25.09459	33.44966	22.9464	22.4804	20.83353	24.79742				
AvgY-	17.10638	20.13318	11.77812	22.28138	22.74738	24.39425	20.43036				
Effect	11.01501	4.961412	21.67154	0.665018	-0.26698	-3.56072	4.367055				
Var+	7.685635	11.14461	7.159463	6.260733	1.953159	5.476117	0.538815				
Var-	5.664244	2.20527	6.190417	7.089146	11.39672	7.873763	12.81106				
F	1.356869	5.053626	1.15654	0.883143	0.171379	0.695489	0.042059				



Minitab DOE

Minitab can be used to create the framework for this experiment. Below minitab has created a an experiment where we have our 3 factors, our 2 levels and our 16 runs. Note that minitab has randomized the order in which each run is take place. In addition we use coded values in minitab as well.

	C1	C2	C3	C4	C5	C6	C7	C8
	StdOrder	RunOrder	CenterPt	Blocks	time	temp	catalyst(AorB)	cost
2	7	2	1	1	-1	1	1	16.3465
3	10	3	1	1	1	-1	-1	8.0205
4	9	4	1	1	-1	-1	-1	-0.4931
5	15	5	1	1	-1	1	1	9.9135
6	4	6	1	1	1	1	-1	6.5004
7	2	7	1	1	1	-1	-1	6.9885
8	1	8	1	1	-1	-1	-1	0.5182
9	12	9	1	1	1	1	-1	13.3288
10	8	10	1	1	1	1	1	20.7294
11	11	11	1	1	-1	1	-1	6.5743
12	16	12	1	1	1	1	1	21.3106
13	3	13	1	1	-1	1	-1	5.6749
14	6	14	1	1	1	-1	1	15.9695
15	14	15	1	1	1	-1	1	19.6377
16	5	16	1	1	-1	-1	1	15.6700

Minitab DOE

The below minitab analysis of the DOE shows each of the three effects as well as the interactions (combinations) of effects.

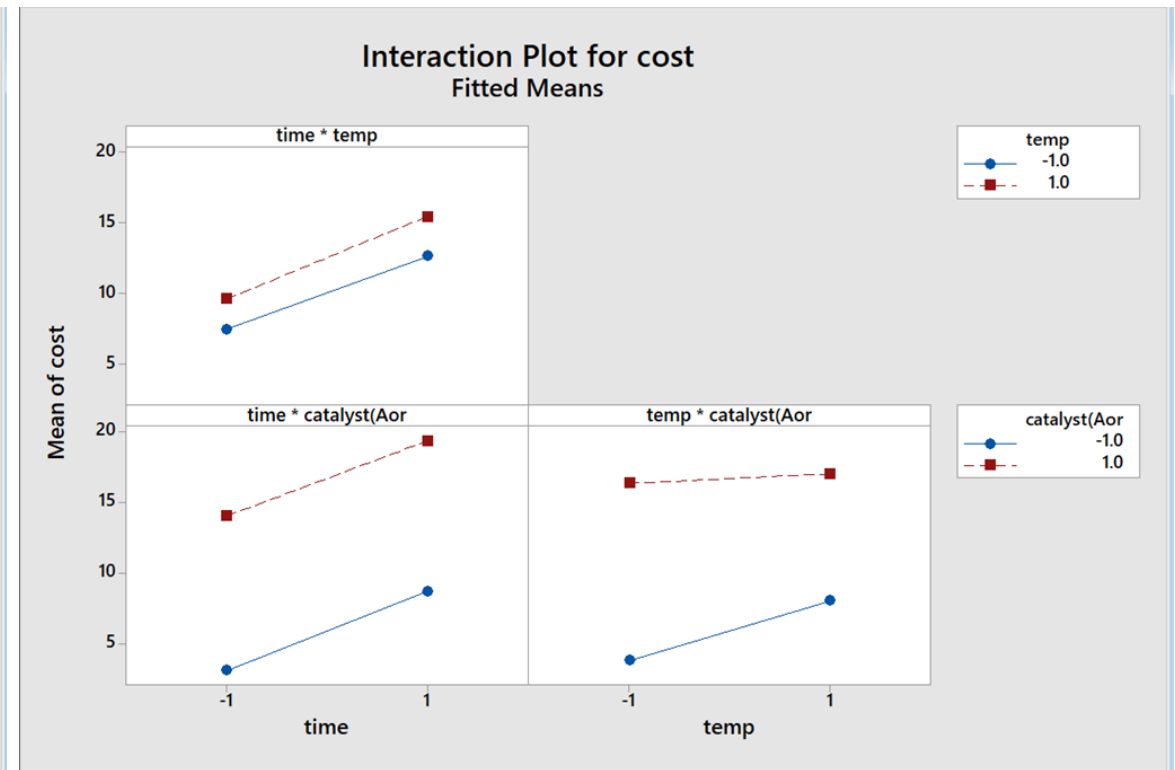
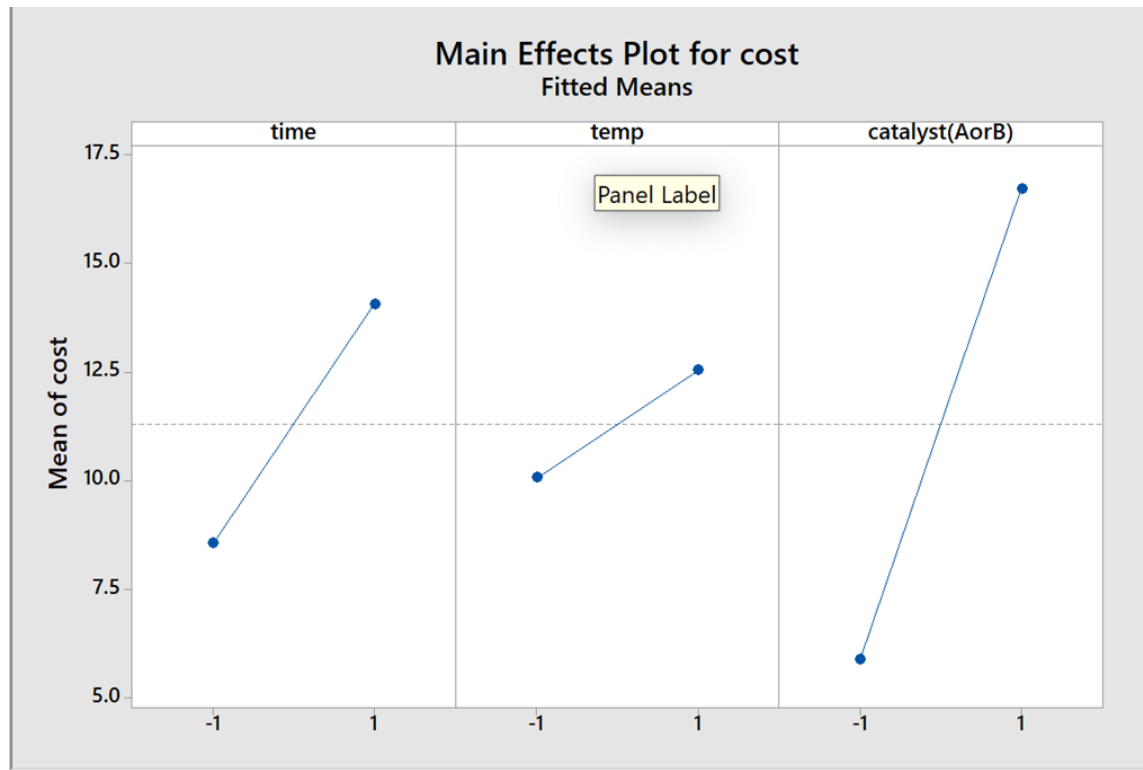
Term	Effect	Coef	SE Coef	T-Value	P-Value	VIF
Constant		11.307	0.646	17.51	0.000	
time	5.508	2.754	0.646	4.26	0.003	1.00
temp	2.481	1.240	0.646	1.92	0.091	1.00
catalyst(AorB)	10.836	5.418	0.646	8.39	0.000	1.00
time*temp	0.333	0.166	0.646	0.26	0.803	1.00
time*catalyst(AorB)	-0.133	-0.067	0.646	-0.10	0.920	1.00
temp*catalyst(AorB)	-1.780	-0.890	0.646	-1.38	0.205	1.00
time*temp*catalyst(AorB)	2.184	1.092	0.646	1.69	0.129	1.00

Factor	Has Effect?
Time	YES
Temp	NO
Catalyst	YES
Time * Temp	NO
Time * Catalyst	NO
Temp * Catalyst	NO
Time*Temp*Catalyst	NO

The interaction between Time and Temp (Time*Temp) has a p-value of 0.589 which is above our alpha value of .05 (95% confidence). We make the assessment that the experiment is governed by effects from each factor as well as interactions between some factors.

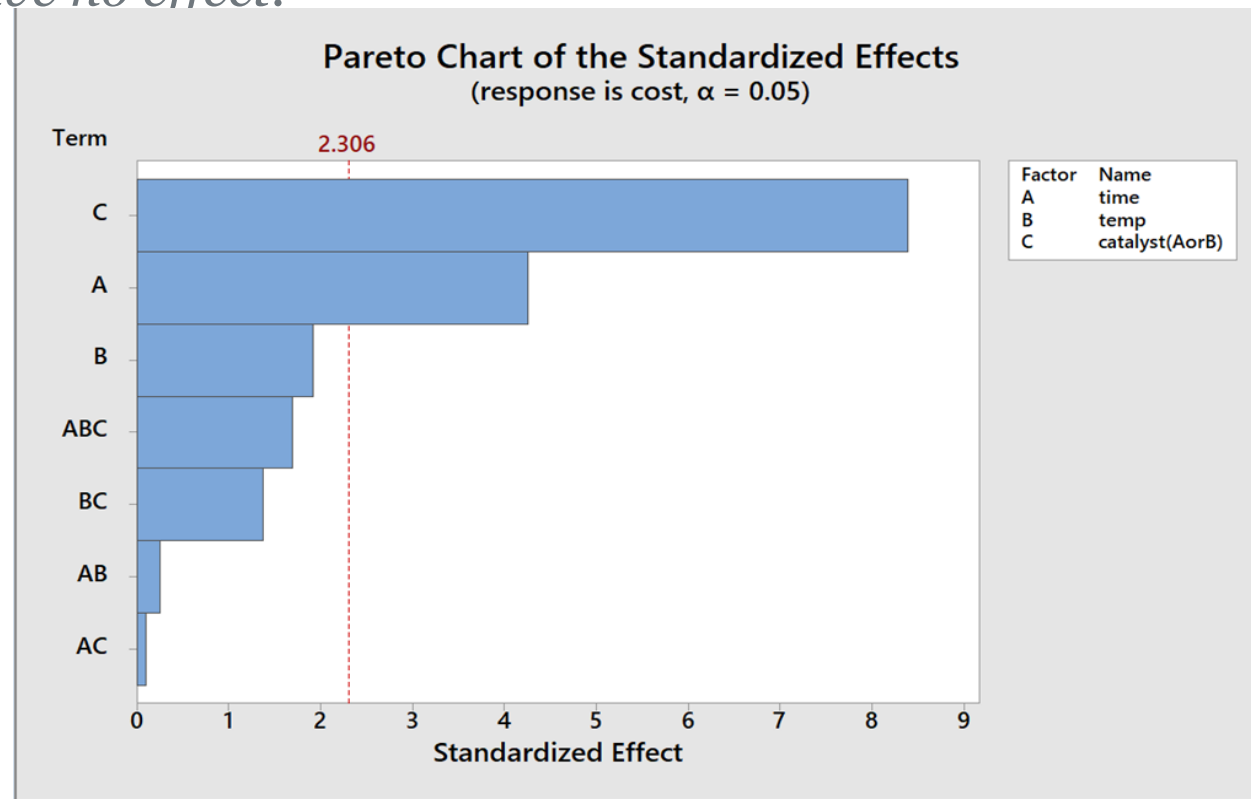
DOE Interpretation using graphs

Minitab graphs can give us the same information as the numerical analysis shown above. Below are graphs for each factor as well each interaction.



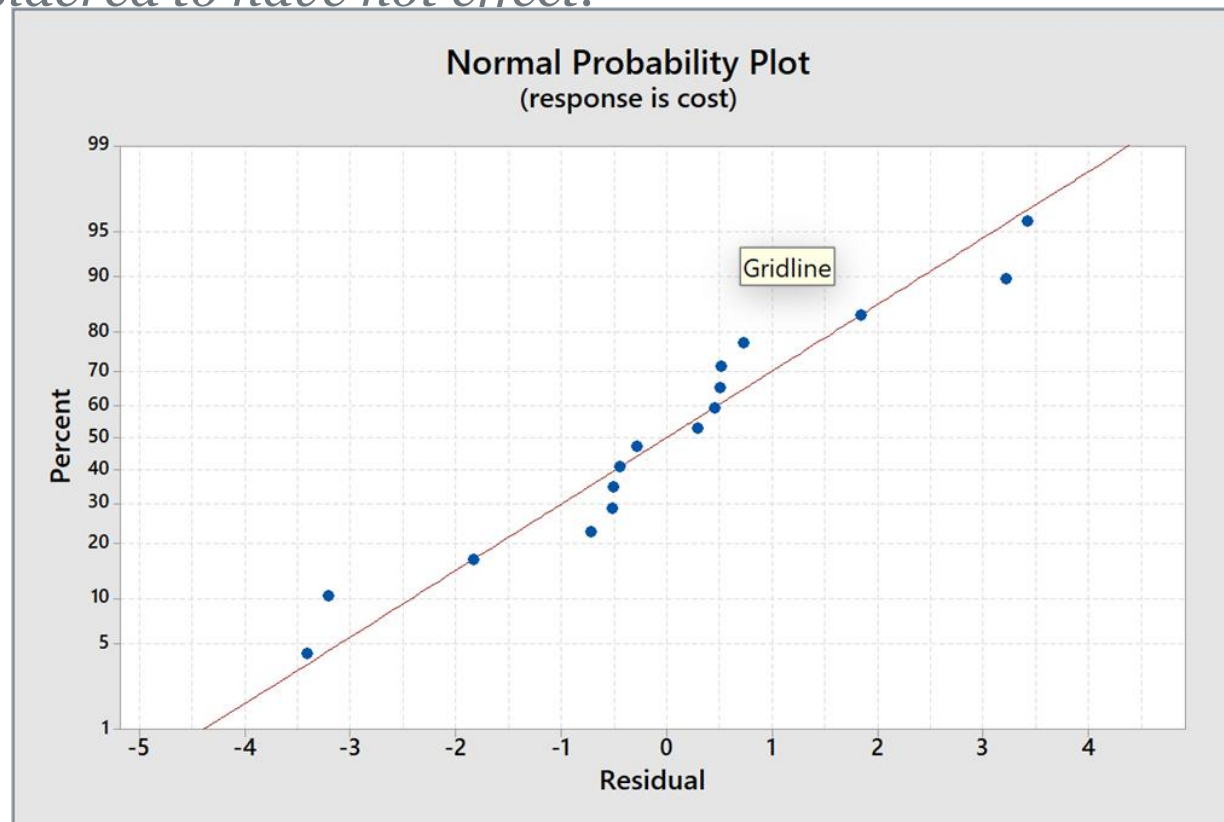
DOE Interpretation using graphs

This graph is a Pareto chart. With the Pareto chart we see a boundary line; this line is a 95% confidence boundary. Factors and interactions that go beyond this line are assumed to have an effect, while factors and interactions that do not pass this line are assumed to have no effect.



DOE Interpretation using graphs

Similarly in the graph to the right in the above figure there is a blue line and several dots representing the factors and interactions. The dots that are red and a distance from the line are considered to have an effect while the dot(s) that are black and near the line are considered to have not effect.



DOE Estimated coefficients

Minitab automatically calculates the constants and coefficients we need for our predictive equation. Similar to what we did in Excel. The below table displays the minitab calculated values for this experiment.

Term	Coef
Constant	11.307
time	2.754
temp	1.240
catalyst(AorB)	5.418
time*temp	0.166
time*catalyst(AorB)	-0.067
temp*catalyst(AorB)	-0.890
time*temp*catalyst(AorB)	1.092

Combine the Constant along with the coefficients to determine the predictive equation.

$$\begin{aligned} \text{cost} = & 11.307 + 2.754 \text{ time} + 1.240 \text{ temp} + 5.418 \text{ catalyst(AorB)} + 0.166 \text{ time*temp} \\ & - 0.067 \text{ time*catalyst(AorB)} - 0.890 \text{ temp*catalyst(AorB)} \\ & + 1.092 \text{ time*temp*catalyst(AorB)} \end{aligned}$$

Reliability

According to our design of the Mudslide Prevention System, it has the following five subsystems:

- (1) monitoring system;*
- (2) infrastructure building team;*
- (3) supply management team;*
- (4) rescue team;*
- (5) communication maintaining group.*

If our Mudslide prevention System requires a Reliability of 95%, we need reliability of 99% to each subsystem:

$$R(IICU) = R(D)*R(N)*R(SS)*R(V)*R(A)E = 0.99^5 = 0.95099$$

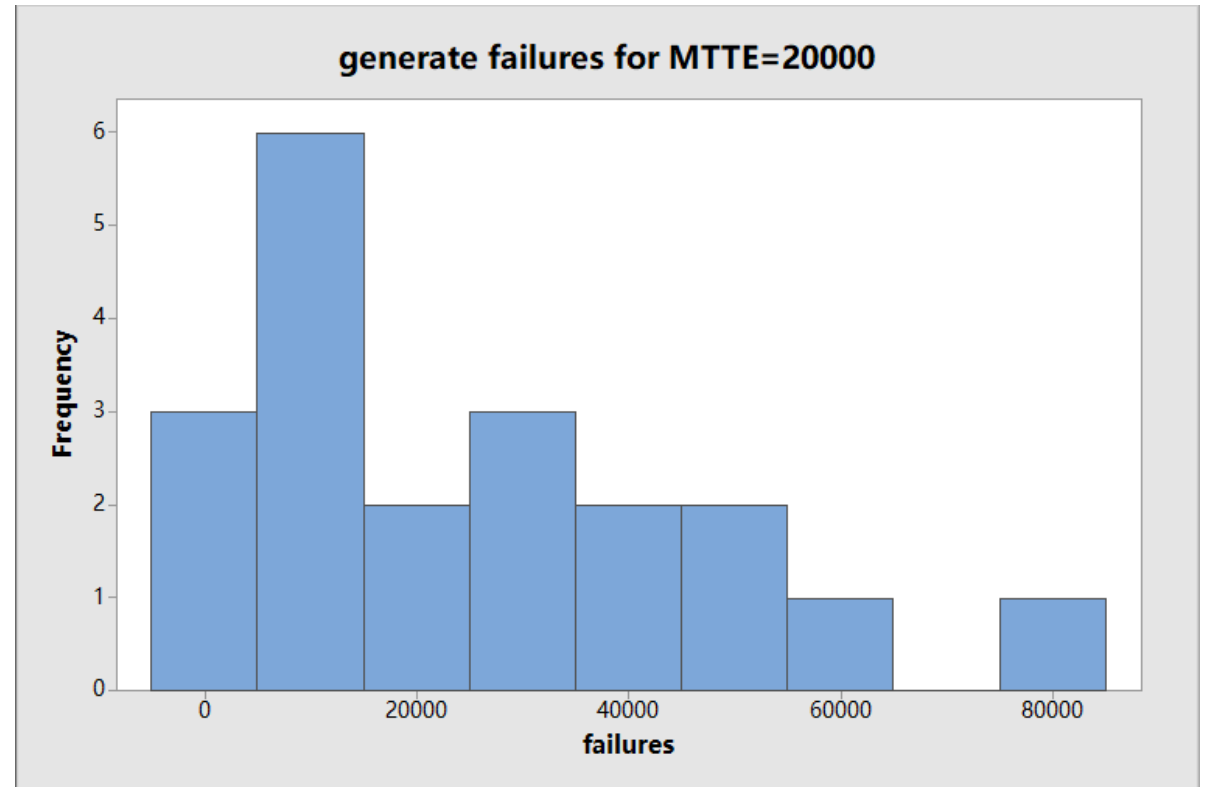
Generate data

	Failures		
1	46462.9	11	17649.7
2	38890.2	12	38423.9
3	8386.1	13	22927.1
4	10016.1	14	6426.1
5	2340.1	15	1392.1
6	33690.7	16	25150.1
7	58067.6	17	6714.9
8	48176.9	18	7339.8
9	34743.3	19	84.3
10	6551.9	20	77648.8
		SUM	491082.6
		AVG	24554.13

Reliability

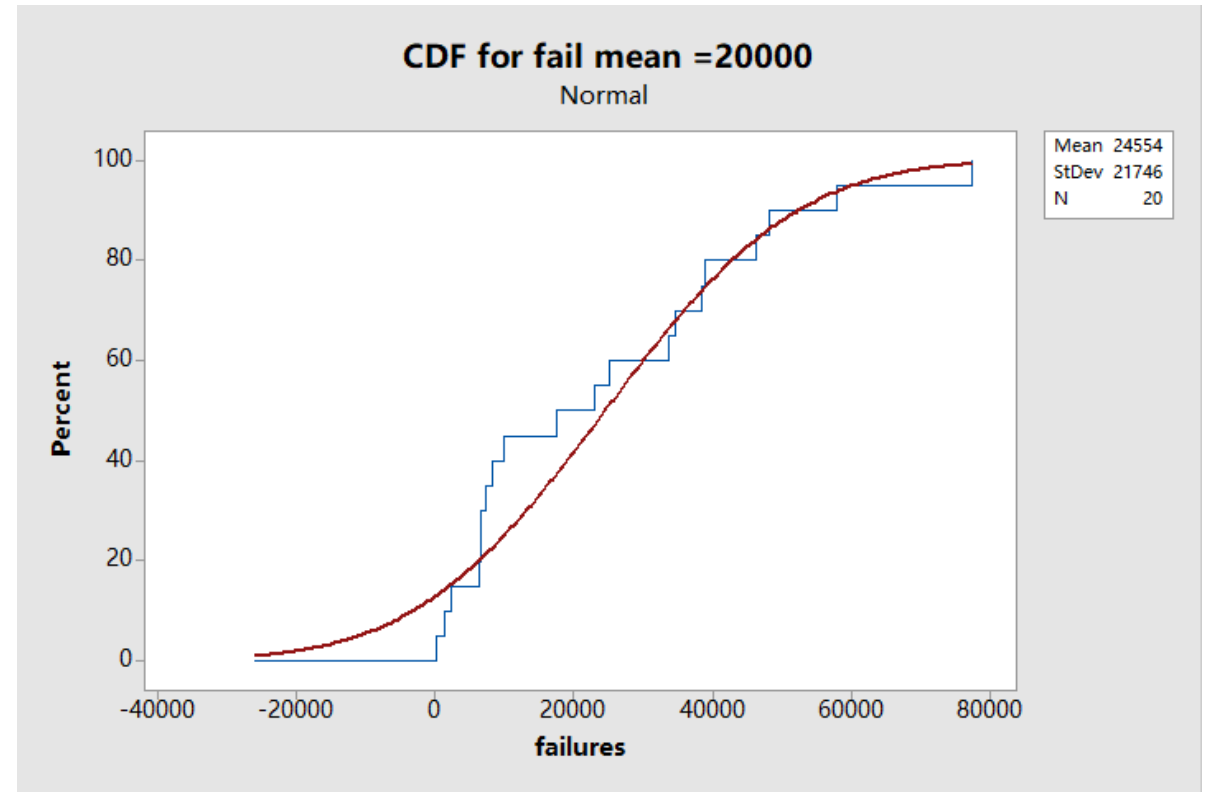
Example of monitoring system: receiving and assessing a lot of $n = 20$ vibration detectors.

Generate the lives (time to failure) of $n=20$ vibration detectors, as Exponential with Mean Time Between Failures: $MTBF = 20K$ hours



Reliability

The diagram below is the cumulative distribution function (CDF). According our data (after sorted), 20% of the failure are less than 6426.1, the last 10% of the failure are greater than 58067.6.



Reliability

If we want to define a Ventilator non-stop work time of 5000 hrs(= 83.3 hrs = 3.5 days):

95% CI for Reliability on this Mission Time (working without stopping for sched maintenance):[0.29 ; 0.61]

Since the upper internal is 0.61, that means there are 40% of the time our vibration dectectors are not working. Such reliability is not acceptable, so we will decrease Mission (maint.) time to 480 min = 8 hours

95% CI for Reliability on this Mission Time : [0.891; 0.953]

Reliability

But we do not have the time nor the resources to wait for the Complete sample of $n = 20$ vibration dectectors to have a failure. We will stop our test at the First Failure $X_k: k = 1$ (assume we stop testing at $time=84.3$)

Then we calculate a new 95% CI of the mean is [457, 66,600]

We have reduced testing time from 77648 to 84 min, but paid a price of a much larger CI.

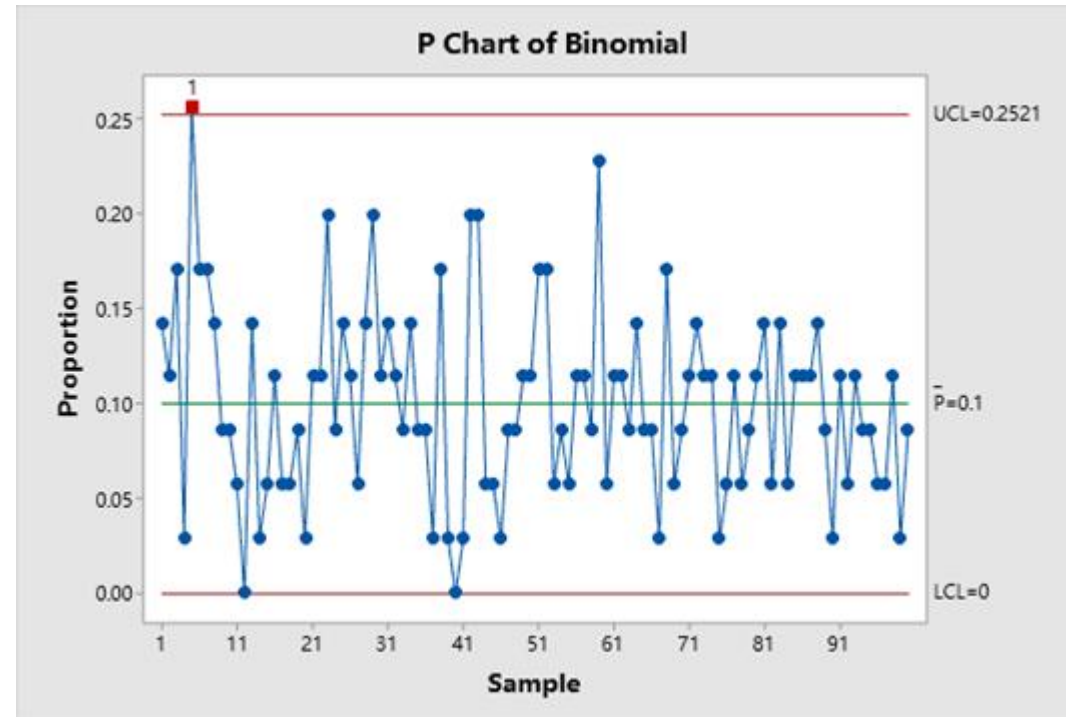
Control

Strategies to control the improvement:

- Document the improved process
- Regular maintenance of facilities built for mudslide
- Making checklist to see if all supplies are available
- Regular evacuation training for citizens

Attribute SPC

One point more than 3.00 standard deviations from center line. Test Failed at points: 5. We can see that percentage is very high at that point, which means the process is unstable. But the mean is going down, and the situation is under control.



Conclusion

We use the methods of Brainstorming , COPQ, fishbone chart, flowchart, let us discover the methods and steps in mitigating mudslide disasters and determine the direction of improvement. Use gage R&R and OC curve to analyze rainfall data, and use VSM and DOE methods to adjust and improve the disaster relief plan to reduce the recovery period. Using reliability analysis reduces equipment testing time and makes the system more reliable. Lastly, we use control chart to control the performance of our system.

Thanks for listening!