

KOREA

DISCOVERY SUMMIT

EXPLORING DATA
INSPIRING INNOVATION



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DOE(Taguchi vs RSM)

Optimization Experimental Design

김광희 (Big Data 분석 전문가)

옥창수 (품질 관리 기술사)

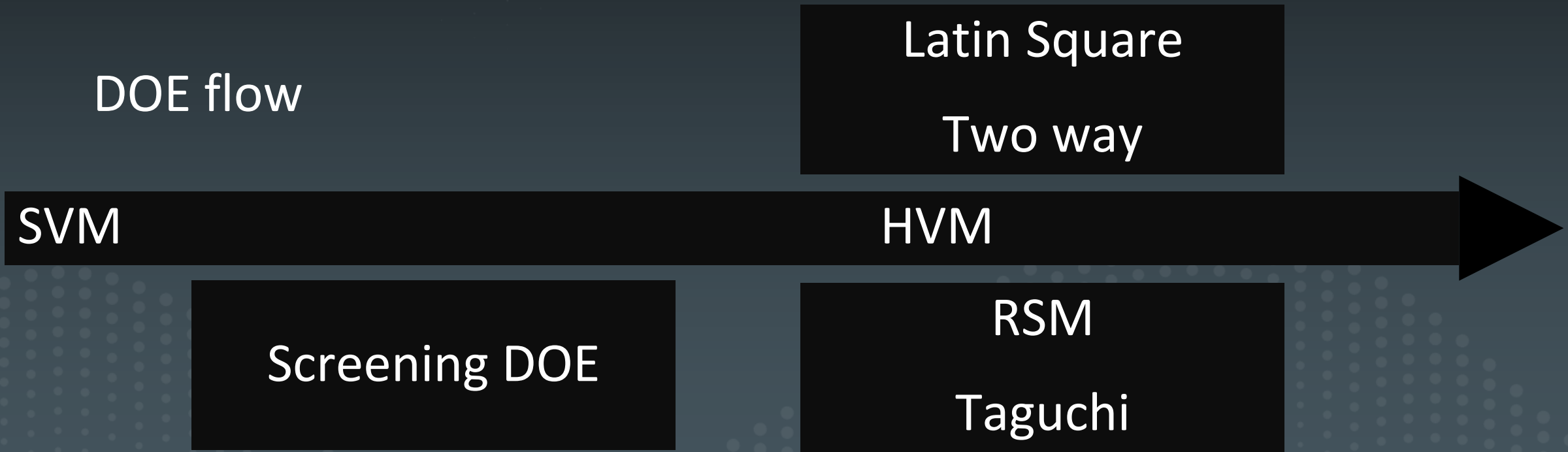
정순우 (품질 관리 기술사)

Content

1. What is DOE
2. DOE simulation
 - Taguchi
 - RSM
3. Result

Background

- Why Taguchi vs RSM ?



DOE 5 principals



Randomization : Random assignment → Minimize bias



Replication : Repeat → Reliability Δ (Degrees of freedom for the error term increase.)



Blocking : Grouping similar units → Controls for variables

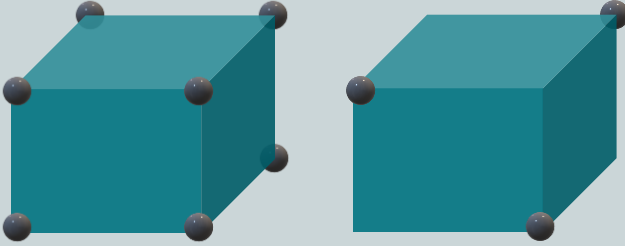
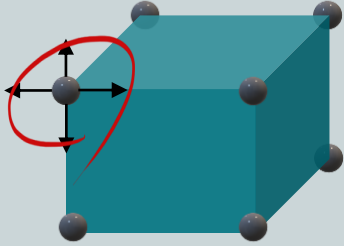
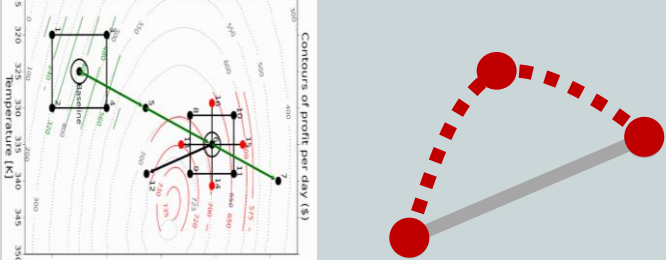


Orthogonality : Independent factor combinations → Reduce confounding



Confounding : Overlap of factor effects → Affects clarity

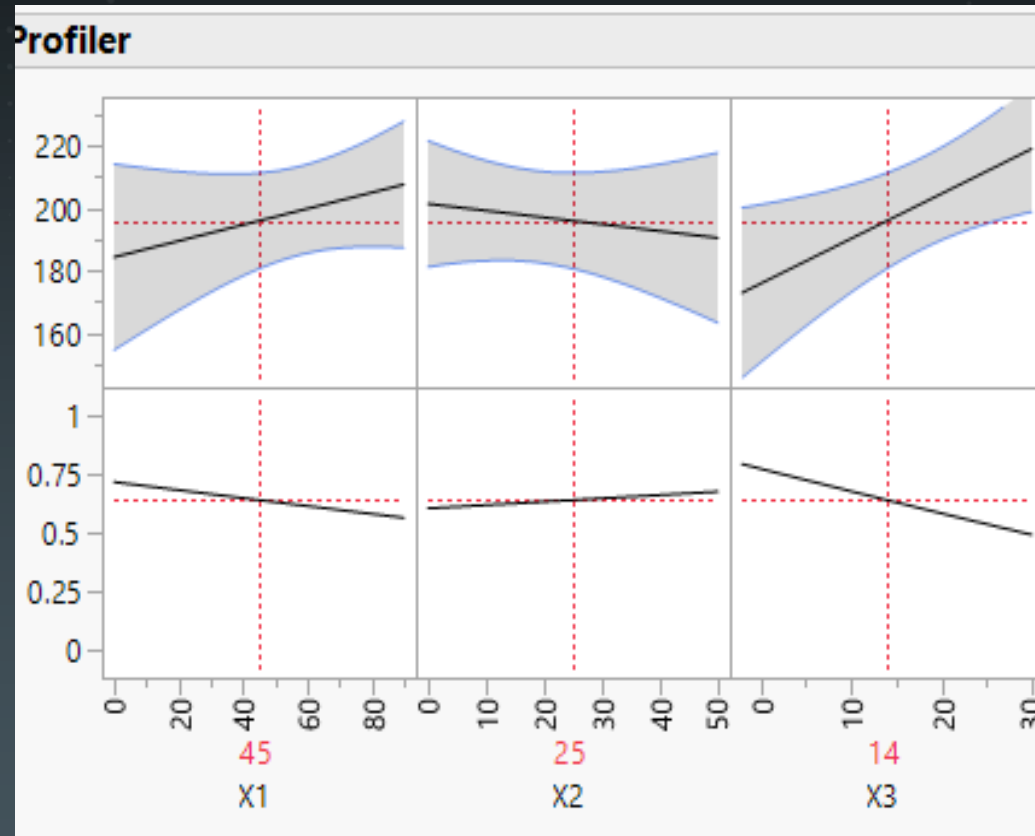
Types of DOE designs

	Factorial Design	Taguchi	RSM
Definition			
Purpose	Study the effect of multiple factors	Minimize variation Robustness design	Identify optimal settings with fewer experiments
Remark	Linear effect	Find loss function	Optimum point can be estimated using curvature

Factorial design

■ Input & Out put

		A1	A2	A3
B1	C1			
	C2			
	C3			
B2	C1			
	C2			
	C3			
B3	C1			
	C2			
	C3			



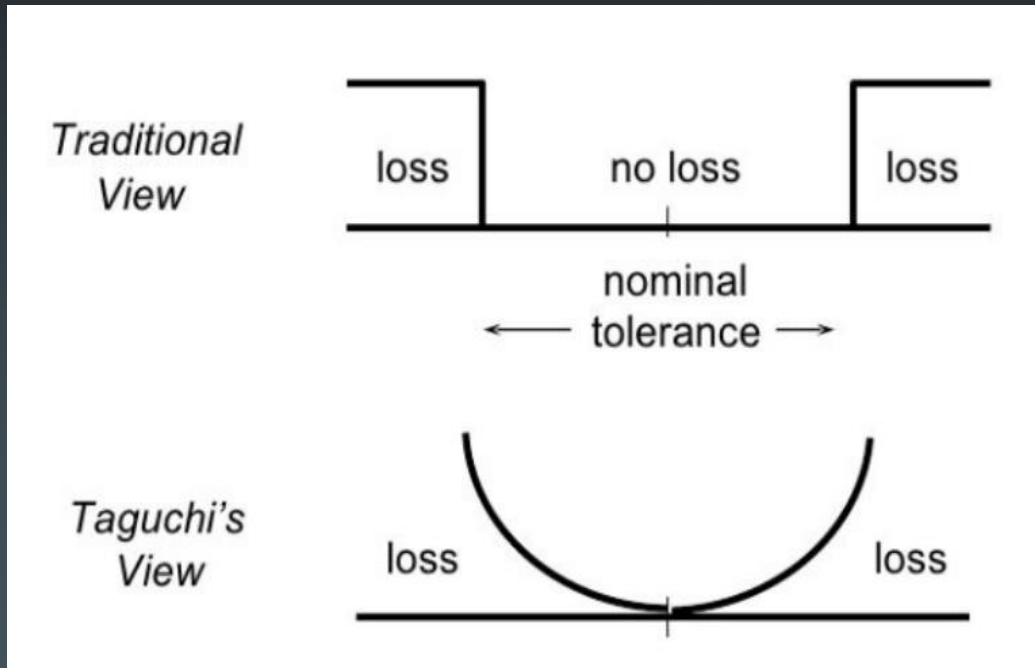
- Cost and Time
- Sensitivity to Noise
- Limited Optimization and Surface Analysis

※ 3^3 : 3-factor, 3-level design

Taguchi design

- Shift → Loss → On target

- Robust design against to noise
→ Preferred higher SNB



- Noise : Uncontrollable factors that affect the experiment
(Ex : Environment condition, User variability and etc.)

Taguchi design

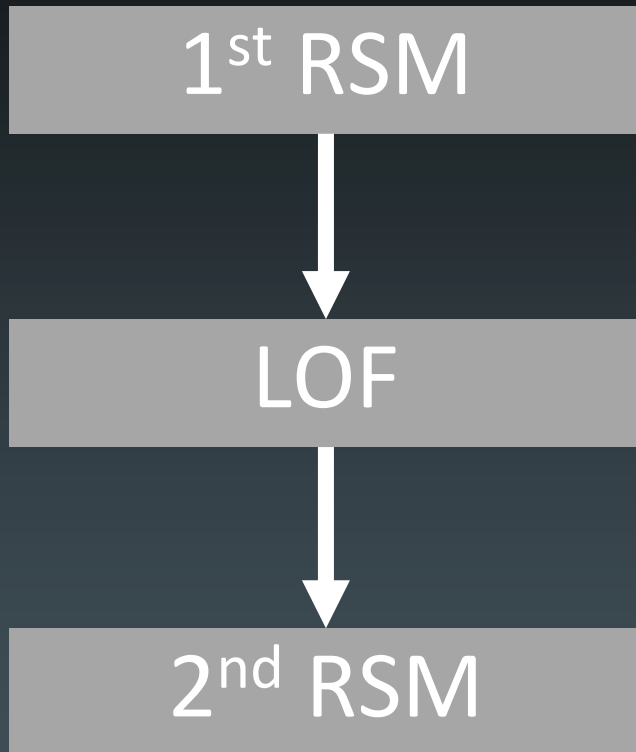
- SNB : Signal to Noise ratio → Preferred higher SNB

- Nominal is Best = $10 \log \left(\frac{\bar{y}^2}{s^2} \right)$  Signal

- Larger is Better = $-10 \log \left(\frac{\sum_{i=1}^n \frac{1}{y^2}}{n} \right)$

- Smaller is Better = $-10 \log \left(\frac{\sum_{i=1}^n y^2}{n} \right)$

RSM design



- 1st order model : Assumes a linear relationship
- 2nd order model : Define optimized condition
 - 3Level : Accounts for a nonlinear relationship

RSM design

※ LOF : Lack of Fit

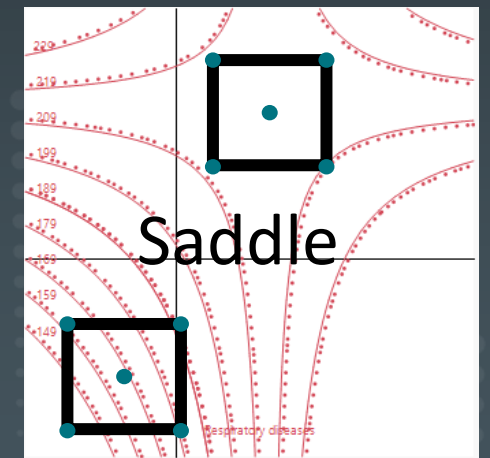
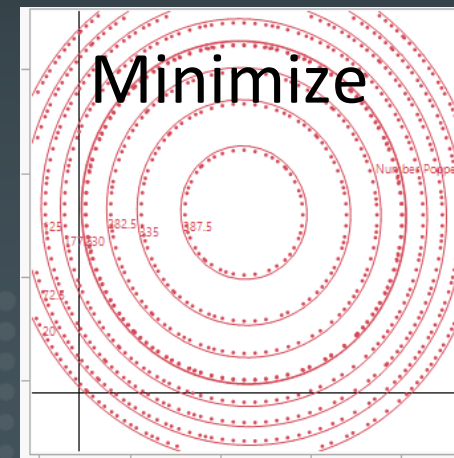
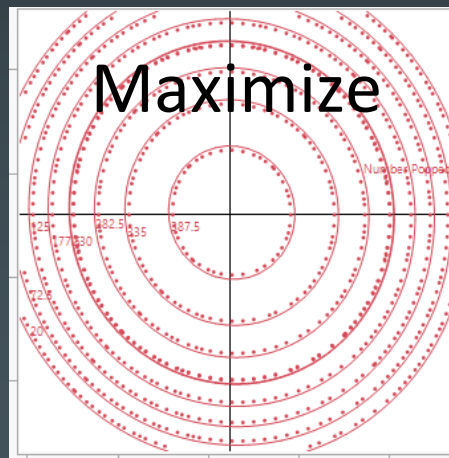
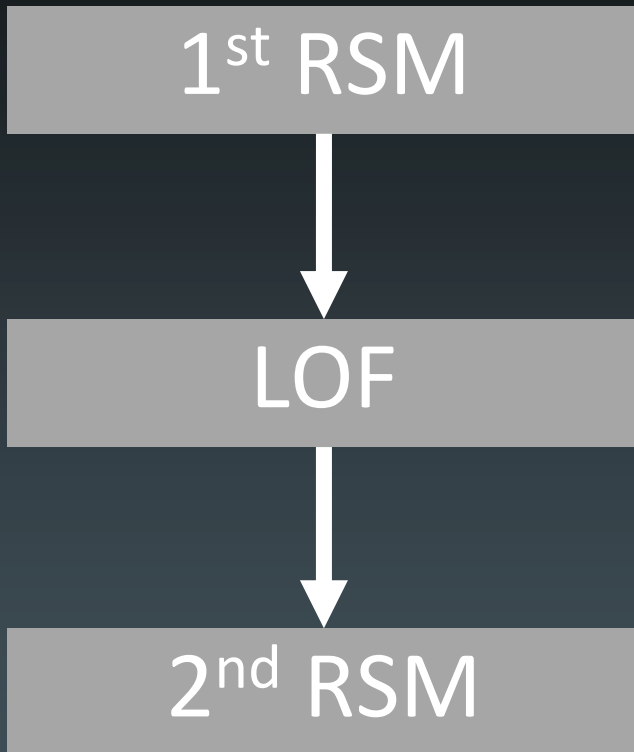
- The assumed regression model is validated through an F-test.

H_0 : No Lack of Fit (LOF).

H_1 : Lack of Fit (LOF) present.

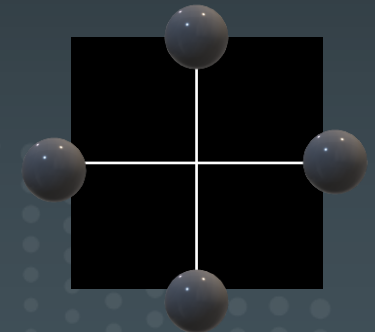
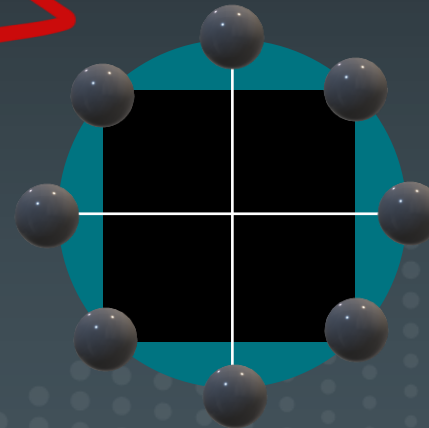
As the F-value increases, and the p-value decreases, H_1 is more likely to be accepted.

※ Contour plot (3types)



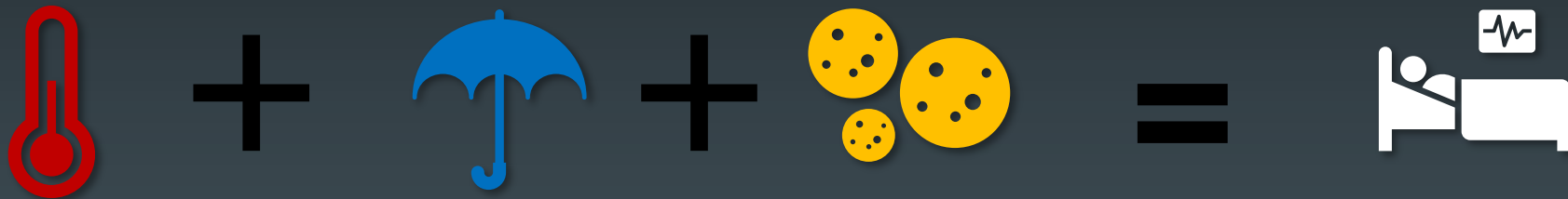
RSM

- CCD : Central composite Design
 - Rotatability : Equal variance
 - Corner : Extreme value
- Box benken
 - Edge : Realistic value



Condition

- Under what conditions of **temperature**, **precipitation**, and **fine dust levels** does the rate of respiratory diseases increase?



- Data reference
 - <https://www.weather.go.kr> (기상청)
 - <https://ncv.kdca.go.kr> (감염병 포털)

Taguchi

DOE - Taguchi Arrays - JMP

File Edit Tables Rows Cols DOE Analyze Graph Tools View Window Help

Taguchi Design

Response

Response Name	Goal	Lower Limit	Upper Limit	Importance
Respiratory diseases	Smaller is Better	145	290	.

Factors

Signal Noise Remove

	Role	Values			Units
2 Level	Signal	1	2	3	
3 Level	Signal	1	2	3	
X3	Signal	1	2	3	
X4	Noise	L1		L2	

- Select Level
- Noise

Taguchi

DOE - Taguchi Arrays - JMP

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Taguchi Design

Response

Response Name	Goal	Lower Limit	Upper Limit	Importance
Respiratory diseases	Smaller is Better	145	290	.

Factors

Name	Role	Values	Units
Precipitation	Signal	0 8.5 150	
Fine dust	Signal	0 20 63	
Temperature	Signal	-2 16 30	
Sex	Noise	Male Female	

4 Factors

Choose Inner and Outer Array Designs

Inner Array

Number	
9	L9 - Taguchi
27	Full Factorial

Continue

Back

- Add Noise
- L9 : L9(3⁴) Frequent orthogonal table in 3level, 4factors

Taguchi

- Mean, SNB (Smaller is Better = $-10 \log \left(\frac{\sum_{i=1}^n y^2}{n} \right)$)

		Precipitation	Fine dust	Temperature	Pattern	-	+	Mean	SN Ratio
1	0	0	-2	---	160	170	165	-44.35366507	
2	0	20	16	-00	181	180	180.5	-45.12957745	
3	0	63	30	-++	218	222	220	-46.84881252	
4	8.5	0	16	0-0	183	185	184	-45.29648474	
5	8.5	20	30	00+	246	242	244	-47.7480883	
6	8.5	63	-2	0+-	165	160	162.5	-44.2180951	
7	150	0	30	+++	226	220	223	-46.96688318	
8	150	20	-2	+0-	180	175	177.5	-44.98482859	
9	150	63	16	++0	186	190	188	-45.48364846	

$A_0 B_2 C_0$

The screenshot shows the JMP software interface for calculating the SN Ratio. The formula bar contains the following expression:

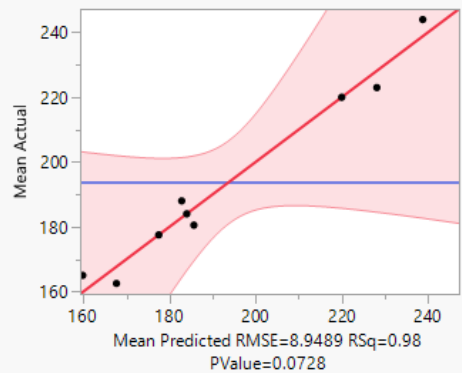
$$-10 \cdot \text{Log}_{10} \left(\text{Mean} \left(\left[- \right]^2, \left[+ \right]^2 \right) \right)$$

The interface includes a filter menu on the left, a list of 8 columns (Precipitation, Fine dust, Temperature, Pattern, -, +, Mean, SN Ratio) in the center, and a toolbar with various mathematical and statistical functions on the right. The 'SN Ratio' column is highlighted in blue.

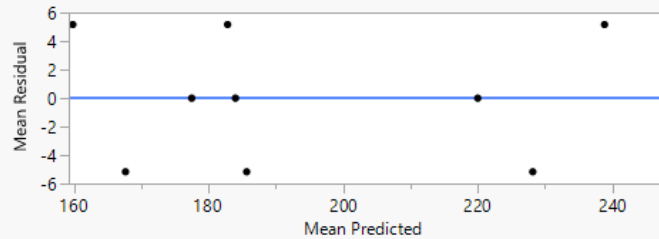
Taguchi

Response Mean

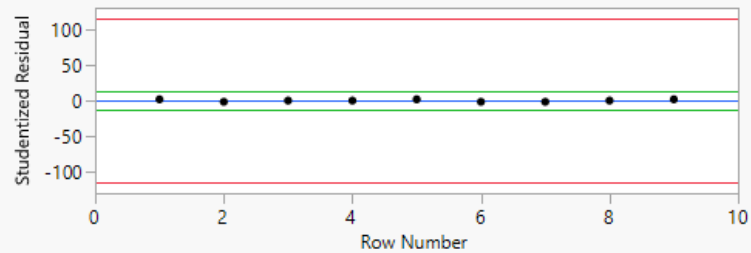
Actual by Predicted Plot



Residual by Predicted Plot



Studentized Residuals



Externally studentized residuals with 95% simultaneous limits (Bonferroni) in red, individual limits in green.

Scaled Estimates

Nominal factors expanded to all levels

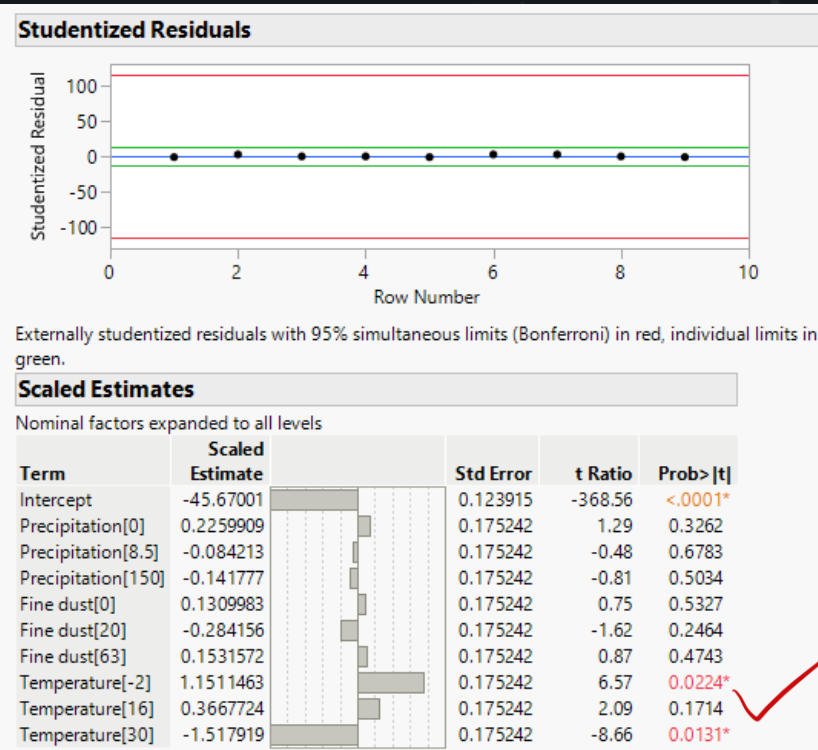
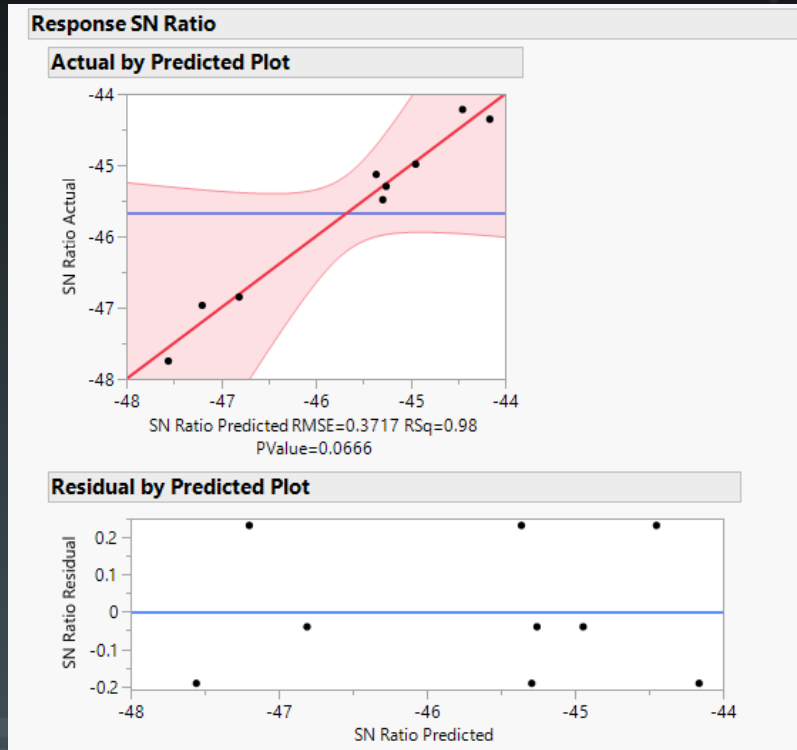
Term	Scaled Estimate	Std Error	t Ratio	Prob> t
Intercept	193.83333	2.982976	64.98	0.0002*
Precipitation[0]	-5.333333	4.218566	-1.26	0.3335
Precipitation[8.5]	3	4.218566	0.71	0.5507
Precipitation[150]	2.3333333	4.218566	0.55	0.6358
Fine dust[0]	-3.166667	4.218566	-0.75	0.5312
Fine dust[20]	6.8333333	4.218566	1.62	0.2467
Fine dust[63]	-3.666667	4.218566	-0.87	0.4764
Temperature[-2]	-25.5	4.218566	-6.04	0.0263*
Temperature[16]	-9.666667	4.218566	-2.29	0.1490
Temperature[30]	35.166667	4.218566	8.34	0.0141*

■ Mean

- C_0 : Mean ↓

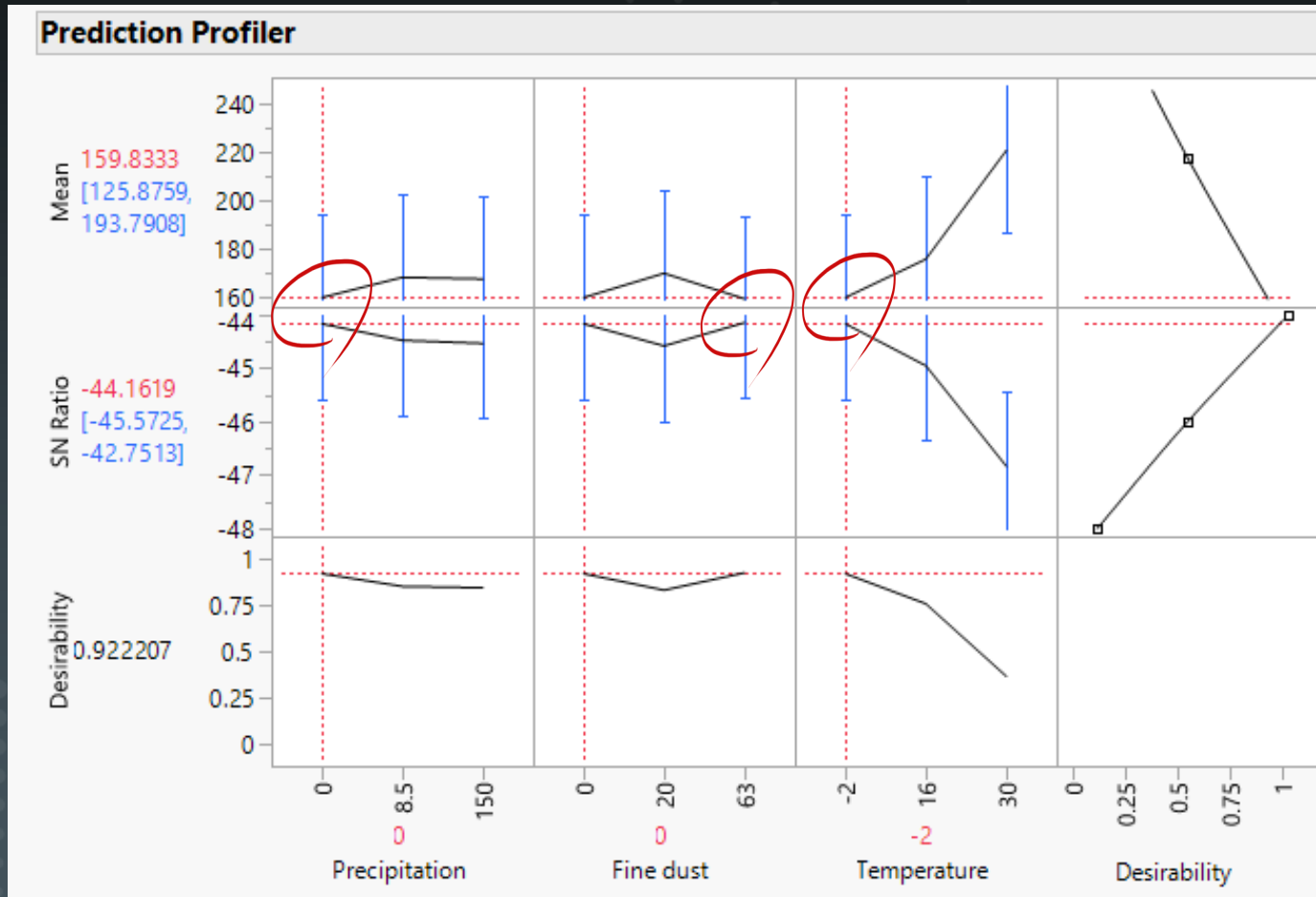
- C_2 : Mean ↑

Taguchi



- SN Ratio
- C_0 : SN Ratio \uparrow
- C_2 : SN Ratio \downarrow

Taguchi



Conclusion

- Mean : $A_0B_2C_0$ ↓

- SN Ratio : $A_0B_2C_0$ ↑

RSM

Response Surface Design

Responses

Add Response Remove Number of Responses...

Response Name	Goal	Lower Limit	Upper Limit	Importance	Units
Respiratory diseases	Minimize	145	290	.	

Factors

Name	Role	Values	Units
Precipitation	Continuous	0 90	
Fine dust	Continuous	0 50	
Temperature	Continuous	-2 30	

3 Factors

Choose a Design

Number of Runs	Block Size	Center Points	Design Type
15		2	Box-Behnken
16		2	Central Composite Design
20		6	CCD-Uniform Precision
20	6	6	CCD-Orthogonal blocks
23		9	CCD-Orthogonal

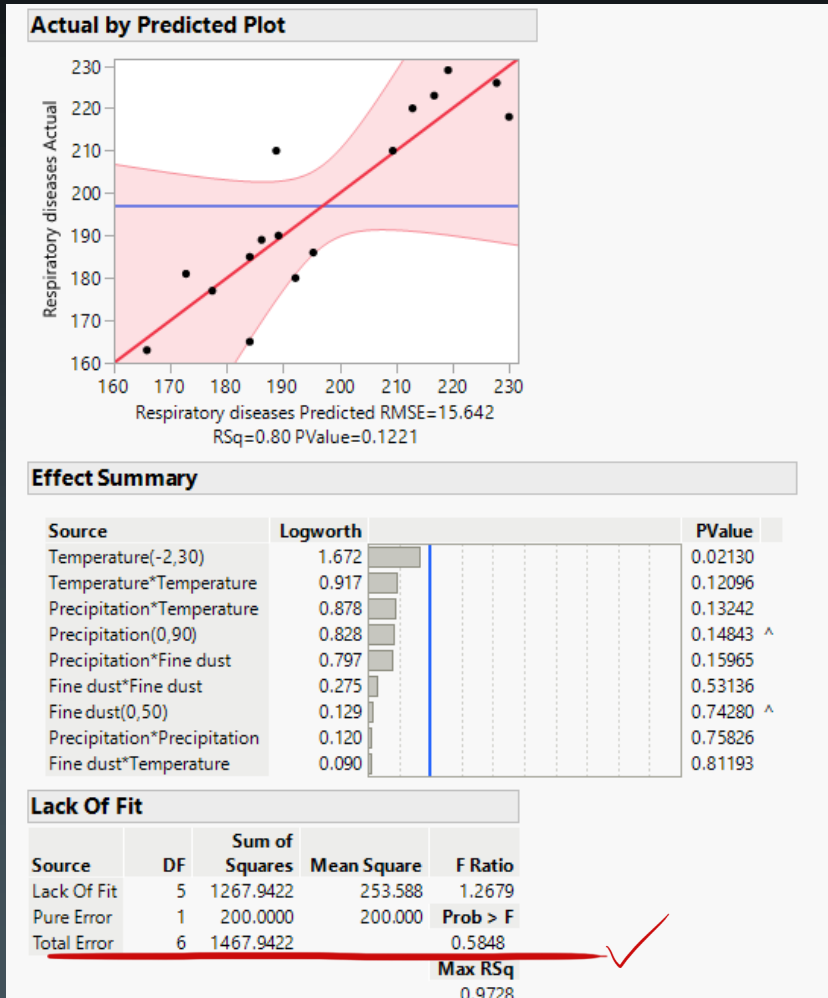
- 2nd RSM (Assume 1st model is not valid)
- Response : 1
- Factors : 3
- Design : CCD

RSM

	▼	▼					
	▼	▼	Pattern	Precipitation	Fine dust	Temperature	Respiratory diseases
1			--++	0	50	30	218
2			0A0	45	50	14	210
3			00a	45	25	-2	149
4			00A	45	25	30	223
5			+---	90	0	-2	229
6			++-	90	50	-2	186
7			+-+	90	0	30	226
8			---	0	0	-2	163
9			A00	90	25	14	190
10			+++	90	50	30	210
11			a00	0	25	14	181
12			0a0	45	0	14	180
13			--+	0	0	30	220
14			000	45	25	14	204
15			-+-	0	50	-2	165
16			000	45	25	14	206

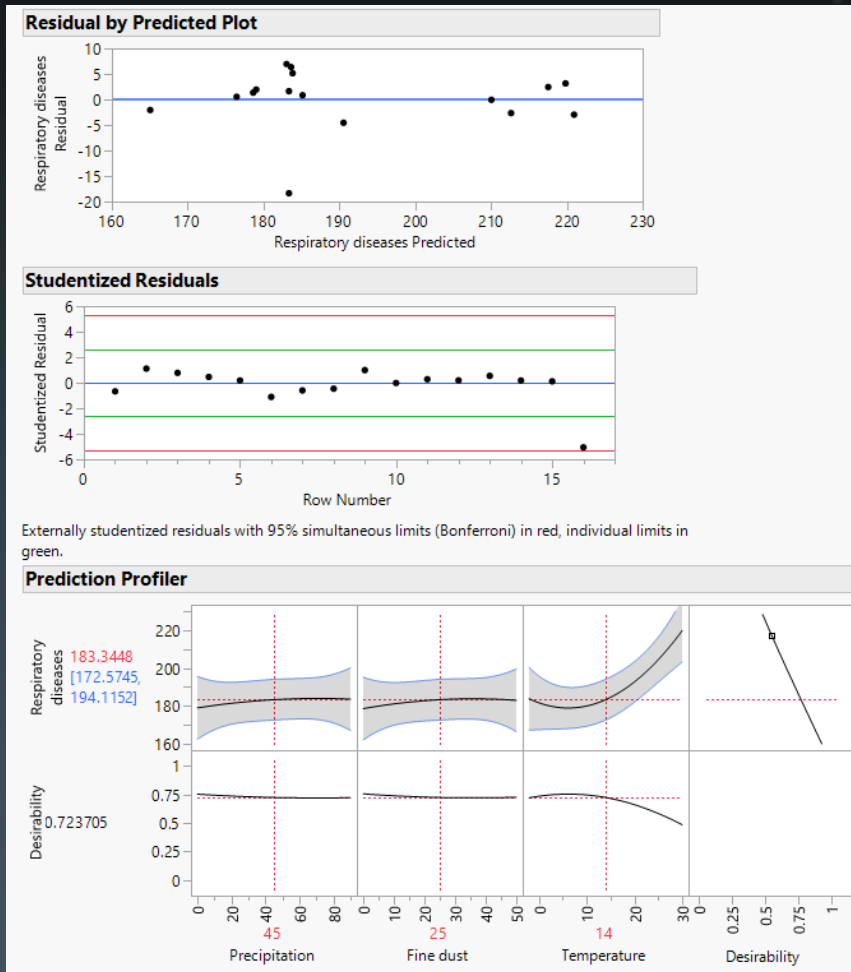
- DOE table
- 16 Legs

RSM



- LOF : P value > 0.05
- ↓
- The lack of fit (LOF) is not significant.
- ↓
- Accept the null hypothesis.
- ↓
- The model is valid.

RSM



- Residuals : Random → No abnormality
- Studentized residual : < 2
- Curvature effect

RSM

Central Composite Design - Fit Least Squares - JMP

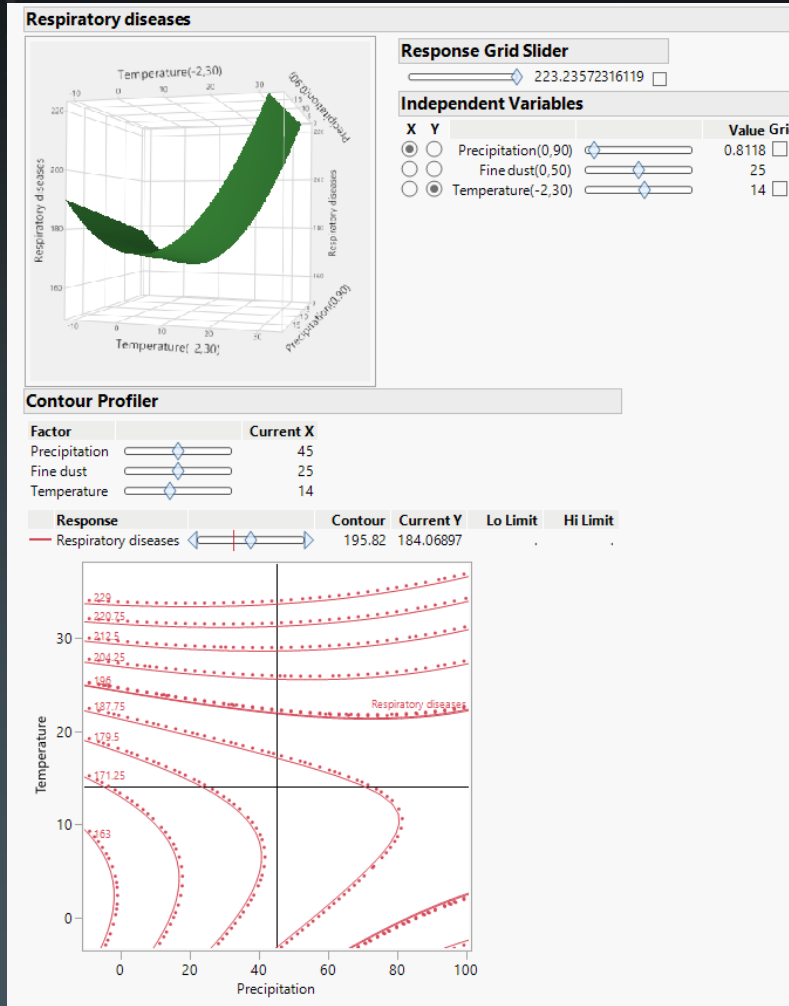
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Response Respiratory diseases

- Regression Reports
- Estimates
- Effect Screening
- Factor Profiling
 - Profiler
 - Interaction Plots
 - Contour Profiler
 - Cube Plots
 - Box Cox Y Transformation
 - Surface Profiler
- Row Diagnostics
- Save Columns
- Multiple Comparisons
- Model Dialog
- Effect Summary
- Local Data Filter
- Redo
- Platform Preferences
- Save Script

Effect Summary

RMSE=15.642
e=0.1221

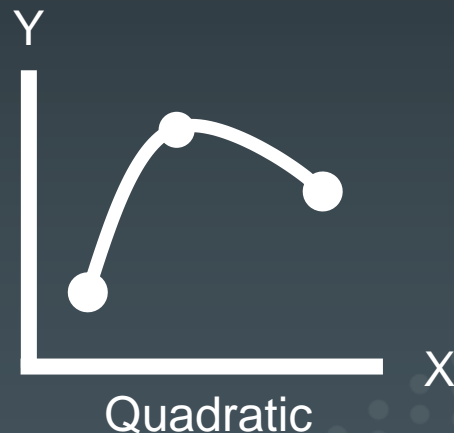


- Surface Profiler shows changes in rate of respiratory diseases

- Contour line shows rate of respiratory diseases

RSM

- If, LOF is not valid \rightarrow Accept null hypothesis \rightarrow Model is valid
- If LOF is valid, need to RSM analysis.



RSM vs Taguchi

	RSM	Taguchi
Purpose	Optimization of response values in the region of interest	Identify optimal conditions that are robust to the effects of noise
Input	Fewer test (using LOF), CCD/BBD	Orthogonal array, Loss function
Output	Response value behavior by input	Robust to noise

Thank you