# Optimization of Polymerization Reactions

**Abstract:** In this research we applied quality engineering principles to a synthetic plastics reaction known as a polymerization reaction, regarded as one of the slowest of all polymerization reactions, and a reaction with a wide molecular mass distribution, to enhance productivity and obtain a polymerized substance with a narrow molecular mass distribution.

## 1. Introduction

In manufacturing processes for chemical products, quality and cost depend strongly on a synthetic reaction condition. The reaction handled in this research, a synthetic reaction of resin called a polymerization reaction, is one of the slowest of all such reactions and has a wide molecular weight distribution. Although we had tried many times to improve productivity and molecular weight distribution, no significant achievement was obtained. In general cases of polymerization reaction, for a complete product we have conventionally measured its molecular weight distribution and analyzed the average molecular weight calculated. The objective of this study was to accomplish an unachieved goal by applying quality engineering to this polymerization reaction. Regarding the application of quality engineering to a chemical reaction, there is one case of doubling productivity by using a dynamic operating window for an organic synthesis reaction. However, the case contains some problems in reproducibility, such as having both peaks and V-shapes in a factor effect plot of the SN ratio.

To solve these problems, we have taken advantage of the ratio of the rate of main reaction  $\beta_1$  and that of side reaction  $\beta_2$ :  $\beta_1/\beta_2$ . By doing this we obtained fairly good reproducibility. In our research, in reference to an organic synthesis reaction, we applied quality engineering to a polymerization reaction by improving a method of handling a polymerization degree distribution and calculating reaction speed  $\beta$  and the SN ratio.

# 2. Experimental Procedure

#### Objective

Since, in general, a polymer substance consists of various types of molecular weight, it has a particular molecular weight. A polymer in this experiment is an intermediate material used as a raw material in the following process. In this case, its molecular weight should be neither too large nor too small. Therefore, using the quality engineering method, we attempted to enhance productivity and obtain a polymerized substance with a narrow molecular weight. To enhance productivity, a high reaction speed and a stable reaction are both essential.

To obtain a polymerized substance with a narrow molecular weight distribution, the constitution of the resulting molecular weight should be high. Therefore, by classifying an ingredient not reaching a certain molecular weight  $M_n$  as an *underreacted substance*, one ranging from  $M_n$  to  $M_m$  as a *target substance*, and one exceeding  $M_m$  as an *overreacted substance*, we proceeded with our analysis.

#### **Experimental Device**

We heated material Y, solvent, and catalyst in a 1-L flask. After a certain time, we added material X to the flask continuously. Next came the reaction of X and Y. When the addition of X was complete, dehydration occurred and polymerization took place.

# Table 1

Control factors and levels for the first  $L_{18}$  orthogonal array (%)

			Level	
Column	Control Factor	1	2	3
1	None	_	—	_
2	A: amount of material X	96	100	106
3	B: amount of agent 1	33	100	300
4	C: amount of agent 2	85	100	115
5	D: reaction speed	90	100	110
6	E: maturation time	0	33	100
7	F: material supply time	0	33	100
8	G: continuously supplied material: X or Y	X	Х	Y

# Table 2

Control factors and levels for the second  $L_{\rm 18}$  orthogonal array (%)

				Level	
Column		Control Factor	1	2	3
1	С:	amount of agent 2	100	115	—
2	<i>A</i> :	amount of material X	92	96	100
3	В:	amount of agent 1	100	300	500
4	Н:	agent type	1	2	3
5	/:	supply method of material	1	2	3
6	J:	supply method of agent 1	1	2	3
7	<i>F</i> :	material supply time	33	100	150
8	К:	material type of Y	1	2	3

# Table 3

Change in constitution of yields at each reaction time (%)

	Reaction Time								
	0.3	1	2	3	4				
Underreacted substance	97.1	30.9	20.7	19.1	12.3				
Target substance	2.9	62.0	72.2	63.6	61.1				
Overreacted substance	0	7.1	7.1	17.3	26.7				
Total	100	100	100	100	100				

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Figure 1 Dynamic operating window method

#### Layout of Orthogonal Array

In the first experiment, we selected seven control factors and allocated them to an  $L_{18}$  orthogonal array with a larger interval of level. Table 1 shows the factor and levels. After analyzing the experimental result of the first  $L_{18}$  experiment, we designed and conducted the second  $L_{18}$  experiment (Table 2).

#### Data Gathering for Analysis

We sampled the polymerizing solution in each of runs 1 to 18 allocated in the  $L_{18}$  orthogonal array at a certain interval. Then, using gel permeation chromatography (GPC), we gauged the constitution of a material and polymers.

As a specific example, Table 3 shows a change in the constitution at each reaction time in run 2 in the first  $L_{18}$  orthogonal array. The data for a molec-



Figure 2 Reaction regarded as ideal

ular mass and its distribution were computed as converted values of the standard polystyrene molecular mass.

# Analysis by Dynamic Operating Window Method

#### Generic Function

Since a chemical reaction is an impact phenomenon between molecules, by setting the constitution of underreacted substances in a yielded substance to p, we defined the following equation as a generic function on the basis that reaction speed is proportional to density of material:

$$b = e^{-\beta T}$$

Taking the reciprocal of both sides and logarithmizing, we have

$$\ln(1/p) = \beta T$$

Setting  $Y = \ln (1/p)$ , we obtain the following zeropoint proportional equation:

$$Y = \beta T$$

Now  $\beta$  is a coefficient of the reaction speed.

#### Concept of Operating Window

If we categorize a yield as underreacted, target, and overreacted based on the molecular mass, as a chemical reaction progresses, the constitution of the target substance rises (Figure 1).

Setting the constitution of an underreacted substance at the point of reaction time *T* to *p* and that of an overreacted substance to *q*, we can express that of an overreacted substance by 1 - p - q. Since an ideal reaction is regarded as one with a faster reaction speed and higher constitution then a target substance (Figure 2), we should lower the constitution of an underreacted substance immediately and maintain that of an overreacted substance as low as possible.

Then, taking into account the following two reaction stages, we estimated a reaction speed for each stage  $\beta$ .

 $R_1$ : a material (molecular mass  $M_1$ , monomer) and polymers (molecular mass  $M_2 - M_{n-1}$ , polymer, that is, a dimer consisting of two combined molecules) convert into target substances (molecular mass  $M_n - M_{m-1}$ ) and overreacted substances (molecular mass more than  $M_m$ ):

material + polymer  $(M_2)$  + …

+ polymer  $(M_{n-1})$   $(M_n + \dots + M_{m-1})$  +  $M_m \dots$ 

We define the reaction speed at this stage as  $\beta_1$ .

 $R_2$ : A material (molecular mass  $M_1$ , monomer), polymers (molecular mass  $M_2 - M_{n-1}$ , polymer, that is, a dimer consisting of two combined molecules), and target substances (molecular mass  $M_n - M_{m-1}$ , monomer) convert into overreacted substances (molecular mass more than  $M_m$ ):

material + polymer 
$$(M_2)$$
 +  $\cdots$   
+ polymer  $(M_{n-1})$  +  $(M_n$  +  $\cdots$  +  $M_{m-1})$   $M_m$   $\cdots$ 

We define the reaction speed at this stage as  $\beta_2$ .

A reaction speed in each stage is not a theoretical rate of a simple chemical reaction per se, however, but a rate in which the underreacted substances described above convert into a yield. The reason that we deal with the reaction in this manner is to prepare a mechanism of additivity. As a result, since  $R_1$  indicates a yielding rate of target and overreacted substances and  $R_2$  represents that of an overreacted substance, our final objective is to obtain an optimal configuration to maximize the difference between two rates,  $R_1$  and  $R_2$ .

Now for  $R_1$  and  $R_2$  we calculate the constitution of an underreacted substance at the point of time  $T_i$ ,  $p_i$  and  $p_i + q_i$ , and compute rates of reaction  $\beta_{1i}$ and  $\beta_{2i}$ .

$$R_{1}: \quad p_{i} = e^{-\beta_{1i}/T_{i}} \qquad \ln \frac{1}{p_{i}} = \beta_{1i}T_{i}$$

$$R_{2}: \quad p_{i} + q_{i} = e^{-\beta_{2i}T_{i}} \qquad \ln \frac{1}{p_{i} + q_{i}} = \beta_{2i}T_{i}$$

In this case, the reaction speed to yield a target substance  $\beta_1$  should be larger and the reaction speed to yield an overreacted substance  $\beta_2$  should be smaller.

$$\beta_{1i} = \frac{Y_{1i}}{T_i} = \frac{1}{T_i} \ln \frac{1}{p_i}$$
$$\beta_{2i} = \frac{Y_{2i}}{T_i} = \frac{1}{T_i} \ln \frac{1}{p_i + q_i}$$

Using these equations, we can calculate  $\beta_{1i}$  and  $\beta_{2i}$ for each of *K* measurement points. Now, since we need to maximize a reaction speed of yielding a target substance,  $\beta_{1i}$  and to minimize that for an overreacted substance,  $\beta_{2i}$ , by computing  $\beta_{1i}$  as a larger-the-better SN ratio  $\eta_1$  and  $\beta_{2i}$  as a smaller-thebetter SN ratio  $\eta_2$ , we find factors to maximize  $\eta_1$ +  $\eta_2 = \eta$ . This is a dynamic operating window method.

$$\eta_{1} = -10 \log \left[ \frac{1}{K} \left( \frac{1}{\beta_{11}^{2}} + \dots + \frac{1}{\beta_{1K}^{2}} \right) \right]$$
$$\eta_{2} = -10 \log \left[ \frac{1}{K} (\beta_{21}^{2} + \dots + \beta_{2K}^{2}) \right]$$

When a reaction speed is stable throughout a total reaction, we can analyze the reaction with only one  $\beta$  at a single observation point in a more efficient manner. If we use a single observation point, the calculation is

$$\begin{split} \eta &= \eta_1 + \eta_2 \\ &= -10 \, \log \frac{1}{\beta_1^2} - 10 \, \log \beta_2^2 = 10 \, \log \frac{\beta_1^2}{\beta_2^2} \end{split}$$

As sensitivity *S*, we consider  $\beta_1$  regarding productivity:

$$S = 10 \log \beta_1^2$$

Example of Analysis

We now show the calculations for the experimental data for run 2 in the first  $L_{18}$  orthogonal array. The observation points of  $\beta_1$  and  $\beta_2$  can differ. Then, setting the observation points so as to estimate  $\beta_1$  and  $\beta_2$  most accurately at  $T_1$  and  $T_2$ , respectively, we compute  $\beta_1$  and  $\beta_2$  from each single point.

Decreasing speed of an underreacted substance:

$$\beta_1 = \frac{1}{T_1} \ln \frac{1}{p_1} = \frac{1}{1} \ln \frac{1}{0.309} = 1.1757$$



Factor effect plot of the SN ratio in the first  $L_{18}$  experiment

Yielding speed of an overreacted substance:

$$\beta_2 = \frac{1}{T_2} \ln \frac{1}{p_2 + q_2} = \frac{1}{4} \ln \frac{1}{0.123 + 0.611} = 0.0776$$

SN ratio:

$$\eta = 10 \log \frac{\beta_1^2}{\beta_2^2} = 10 \log \frac{(1.1757)^2}{(0.0776)^2} = 23.6105 \text{ dB}$$

Sensitivity:

$$S = 10 \log \beta_1^2 = 10 \log (1.1757)^2 = 1.4059 \text{ dB}$$

## 4. Result of the First $L_{18}$ Experiment

Figures 3 and 4 show the factor effect plots of the SN ratio,  $\eta$ , and sensitivity, *S*, in the first  $L_{18}$  experiment. In addition, Table 4 shows the optimal levels for the SN ratio and sensitivity. By prioritizing sensitivity for factors whose optimal SN ratio and sensitivity are different, we determine factor levels in the second  $L_{18}$  experiment.

# 5. Result of the Second $L_{18}$ Experiment

Considering the result of the first  $L_{18}$  experiment, we determined factors and levels for the second  $L_{18}$ experiment as follows. We have the following control factors:

- *E*: We fixed its levels to zero at  $E_1$  by focusing on productivity because the maturation time is zero under the optimal configuration.
- *D*: We fixed its levels to  $D_2$  by selecting the current level because of its small factor effect.
- *G*: We fixed its levels to  $G_1$  because *X* has turned out to be a good choice in terms of continuously supplied material.

For other factors, *A*, *B*, *C*, and *F*, in order to delve further into optimal levels, we changed each level. On the other hand, as new control factors, we deliberately selected *H*, *I*, *J*, and *K*. As for factor *C*, although we had tested as technically wide a range as possible in the first  $L_{18}$  experiment, we included it once again to associate it with the result in the first  $L_{18}$  experiment because the first column emptied in the layout of the second experiment.



**Figure 4** Factor effect plot of sensitivity in the first  $L_{18}$  experiment

#### Table 4

Optimal configuration in the first  $L_{18}$  experiment<sup>a</sup>

		Factor									
	Α	В	С	D	Ε	F	G				
SN ratio	3	3	(3)	(2)	1	3	(1)				
Sensitivity	1	3	3	(2)	(3)	(3)	1				

<sup>a</sup> Parentheses indicate small-effect level.

Figures 5 and 6 summarize the factor effect plots of the SN ratio,  $\eta$ , and sensitivity, *S*, in the second  $L_{18}$  experiment. According to each result for the SN ratio and sensitivity, we determined the optimal levels as shown in Table 5.

#### Results of Confirmatory Experiment

According to the first and second  $L_{18}$  experiments, we obtained less rugged results for the factor effects of the SN ratio and sensitivity. The implied high additivity of the SN ratio is based on reaction speed. Then we conducted a confirmatory experiment under each optimal configuration of the first and second  $L_{18}$  experiments. Tables 6 and 7 show the results of the first and second  $L_{18}$  confirmatory experiments, respectively.

For the first  $L_{18}$  experiment, we conducted a confirmatory experiment under the optimal configuration for each SN ratio and sensitivity value. The result of the first  $L_{18}$  experiment reveals that whereas we obtained fairly good reproducibility in gain for the SN ratio, there was a small gap in gain for the sensitivity. Under the optimal configuration for sensitivity, we can confirm that the gain improved by 5.0 dB, the reaction speed almost doubled, and the reaction time approximately halved. Because sensitivity  $S = 10 \log \beta_1^2$ ,

$$\begin{aligned} \text{Gain} &= 10 \log \beta_{1\text{optimal}}^2 - 10 \log \beta_{1\text{current}}^2 \\ &= 10 \log \frac{\beta_{1\text{optimal}}^2}{\beta_{1\text{current}}^2} \end{aligned}$$

On the other hand, since the gain is 5.0 dB, that is, the real (nondecimal) number is 3.1,

$$\frac{\beta_{1\text{optimal}}^2}{\beta_{1\text{current}}^2} = 3.1$$
$$\frac{\beta_{1\text{optimal}}}{\beta_{1\text{current}}} = 1.8$$

For the second  $L_{18}$  experiment, we conducted a confirmatory experiment under the optimal configuration determined by both the SN ratio and sensitivity. As a result of the second  $L_{18}$  confirmatory experiment, the gain in actual sensitivity is increased to 9.2 dB from that of the first  $L_{18}$  experiment. This is equivalent to a 2.9-fold improvement in reaction speed (i.e., the reaction speed is reduced by two-thirds).

On the other hand, if we compare two molecular mass distributions of polymerized substances under the current configuration (8720 dB) and the optimal configuration (8840 dB) by adjusting the average molecular mass to the target value, we can obtain almost the same result. That is, we cannot determine a way to narrow down the range of molecular mass distribution in this research.



**Figure 5** Factor effect plot of the SN ratio in the second  $L_{18}$  experiment



Factor effect plot of the sensitivity in the second  $L_{18}$  experiment

## 7. Conclusions

#### Experimental Achievements

The productivity improved dramatically because the reaction speed doubled under the optimal configuration compared to that under the current configuration. It was difficult to improve the molecular mass distribution; consequently, although the reaction speed was enhanced, the molecular mass distribution was not. However, improving the productivity threefold while retaining the same level of current quality (molecular mass distribution) was regarded as a major achievement.

#### Future Reearch

Since the reaction speed was increased despite little improvement in molecular mass distribution, our research can be regarded as a significant success from a company-wide standpoint. However, judging from the definition of SN ratio used, we expected that the molecular mass distribution would also be improved. We explain considerations of improvement in molecular mass distribution as follows.

In the  $L_{18}$  experiments and their confirmatory experiments, the reaction speed  $\beta$  for estimation of the SN ratio is calculated by the data measured at the initial stage of reaction. Looking at the experimental data, we find that in the latter half of the experiment, the reaction saturates and deviates considerably downward from  $Y = \beta T$  (particularly  $\beta_1$ ), as shown in Figure 7. This is because a reverse or termination reaction occurred in the polymerization reaction. If we measure and compare molecular mass distributions in the initial stage of reaction, where such reactions are very minimal, we can observe improvement not only in the SN ratio but also in molecular mass distribution. Then the measurement point of molecular mass distribution in the confirmatory experiment is considered too late.

However, since the initial stage of reaction has small molecular masses as a whole, we need to

# Table 5

Optimal configuration in the second  $L_{18}$  experiment<sup>a</sup>

	Factor										
	С	Α	В	Н	1	J	F	K			
SN ratio	2	1	3	(3)	3	1	2	2			
Sensitivity	2	(2)	2,3	(3)	2,3	1,2	(2)	2			
Optimal level	2	1	3	3	3	1	2	2			

<sup>a</sup> Parentheses indicate small effect level.



**Figure 7** Example of lapse of reaction time

# Table 6

								SN	Ratio	Sensitivity		
Configuration			F	actor	S			Estimation	Confirmation	Estimation	Confirmation	
Optimal η	$A_3$	$B_{3}$	<i>C</i> <sub>3</sub>	$D_3$	$E_1$	F <sub>3</sub>	$G_1$	38.7	34.8	10.8	9.9	
Current	$A_2$	$B_2$	<i>C</i> <sub>2</sub>	$D_2$	$E_{3}$	$F_{3}$	$G_1$	30.6	28.1	3.0	7.2	
Gain								8.1	6.7	7.5	2.7	
Optimal S	$A_1$	$B_{3}$	<i>C</i> <sub>3</sub>	$D_2$	E <sub>3</sub>	F <sub>3</sub>	$G_1$	30.4	29.6	16.3	12.2	
Current	$A_2$	$B_2$	$C_2$	$D_2$	E <sub>3</sub>	F <sub>3</sub>	$G_1$	30.6	28.1	3.0	7.2	
Gain								-0.2	1.5	13.3	5.0	

Results of the first  $L_{18}$  confirmatory experiment

## Table 7

Results of the second  $L_{18}$  confirmatory experiment

									SN	Ratio	Sensitivity		
Configuration			Fa	actor	s				Estimation	Confirmation	Estimation	Confirmation	
Optimal η	$C_2$	$A_1$	$B_{3}$	$H_{3}$	$I_3$	$J_1$	$F_2$	$K_2$	33.1	35.5	19.2	16.4	
Current	$C_1$	$A_3$	$B_1$	$H_2$	$I_1$	$J_{1}$	$F_2$	<i>K</i> <sub>2</sub>	23.0	28.1	5.8	7.2	
Gain									10.1	7.4	13.4	9.2	

continue the polymerization reaction until the average molecular mass attains the target. Our confirmatory experiment demonstrates that after doing so, the molecular mass distribution is no longer improvable. Therefore, to improve molecular mass distribution, research on design of experiments and analytical methods, including consideration of a reverse reaction and consideration of a termination reaction, should be tackled in the future.

### Achievement through Use of Quality Engineering

Experiments in production technology beginning at the beaker level in a laboratory using conventional development processes require a considerable amount of time. By evaluating a polymerization reaction with an SN ratio reflecting a generic function in a chemical reaction, we can improve reproducibility in factor effects and enhance productivity threefold, thereby obtaining a significant achievement in a short period of time.

### References

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