

Optimization of Process Parameters of Plastic Injection Molding Machine by using Taguchi Method

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Abstract: *The present study is application of taguchi method to optimize the process parameters of injection molding machine to minimize the shrinkage of product to enhance the quality of product. The predicted optimal setting ensures minimum shrinkage of product. Five process parameters of plastic injection molding machine were taken as melt temperature, mold temperature, injection pressure, holding pressure and cooling time. By ANOVA analysis it is found that melt temperature is most significant and cooling time is least significant factor affecting the dimensional shrinkage of manufactured plastic product.*

Keywords: *Plastic injection molding, Dimensional shrinkage, Taguchi method, ANOVA Aanalysis, process parameter*

INTRODUCTION AND LITERATURE REVIEW

Plastic injection molding is a technique which is used to produce plastic products. In this technique raw thermo plastic material is used to make the plastic products. Thermo plastic material is having property of easy melting and regaining shape when temperature is lowered. Raw material in form of pellets is fed into hopper. From hopper the raw material is transferred to machine by piston and cylinder mechanism. Raw material is heated to melting temperature with the help of heating elements during movement of raw material to injection chamber. The raw material which is in molten phase is transferred to mold cavity with the help of injection nozzle. The required injection pressure and holding pressure is applied with the help of control panel of plastic injection molding machine. It has the mechanism of product cooling so that product gets solidify. The product is holded for some time for proper cooling of product. After that product was removed from machine using clamping pressure. The product cools fast in mold cavity and finished product is removed from mold cavity. Like metal casting, in PIM also various types of defects are present due to process and design parameters maintained during plastic product manufacturing. Common defects are like shrinkage, warpage, sink mark. Most general and considerable defect is shrinkage of product during PIM process. In small and medium scale industries most of the workers set the process parameter data by their experience so that product dimensional shrinkage can be minimized, they do not have guidance in technical way for proper selection and tuning of process parameters and constants. Proper selection of process parameters in PIM Machine lead to minimize the defects from product and maintain best quality component in first attempt for the specific material [1].

Akbarzadeh, et al. [2] used Polypropylene as raw material and process parameters which were selected for optimization are injection pressure, melt temperature, packing pressure, packing time. Response surface methodology (RSM) is used to optimize the process parameters and response was in the terms of product shrinkage. Different type of regression equations are used to interpolate the data of experiments and to get optimum value of parameters for minimum dimensional shrinkage. Andrisano, et al. [3] used Polyphenylsulfone (ppsu) as raw material and process parameters which were selected for optimization are mold temperature, melt temperature, holding pressure, packing time. Response surface methodology (RSM) and Design of Simulation Experiments (DOSE) is used to optimize the process parameters and response was warpage. Warpage was less affected by mold temperature and more affected by melt temperature & holding pressure. Barghash, et al. [4] used seven process parameters with three levels to reduce shrinkage and warpage of plastic product made of Polypropylene (PP). Taguchi and ANOVA analysis is used to optimize the process parameters. Dimensional shrinkage is more affected by melt temperature and hold time. E. Bociaga, et al [5] used RSM and ANOVA to optimize the process parameters to reduce the cycle time of plastic injection molding process. Lixia, et al. [6] Artificial Neural network (ANN) and Genetic Algorithm (GA) optimization technique is used by researchers for five process parameters to reduce shrinkage defect from plastic cover. Xuan-Phuong Dang, et al., [7] optimized directional shrinkage and war page by using RSM, GA and ANN by taking five factors.

II. EXPERIMENTAL SET-UP

The analysis was done on a plastic injection molding machine with fixed column and moving table as shown in figure 1 named as Electron-155 PIM. The raw material which is used in this research is thermoplastic which will be used to make top cover of food packaging box. This examination was conducted by using design of experiment (DOE) strategy named as Taguchi Method [8]. The essential execution measure in PIM is dimensional shrinkage which demonstrates the quality aspect of plastic product. [9].



Figure 1: PIM machine used for experimentation

Table 1: Process Parameters and their Levels

Factor	Parameter Name	Levels			
		I	II	III	IV
A	Mold Temperature	50	55	60	65
B	Melt Temperature	205	210	215	220
C	Injection Pressure	60	65	70	75
D	Holding-Pressure	70	80	90	100
E	Cooling time	20	25	30	35

Table 1 shows the process parameters and their levels. Total five process parameters and four levels are taken and accordingly orthogonal array has been made.

III. RESULTS AND DISCUSSION

Experiments are carried out by using taguchi L-16 orthogonal array. There is combination of 16 different sets of melt temperature, mold temperature, injection pressure, holding pressure and cooling time which is prepared by taguchi design requirement of five factors and their four levels. The dimensional shrinkage is measured with the help of vernier caliper.

Table 2: Experimental result of Dimensional shrinkage

Sr No	Mold-T	Melt-T	Inj-P	Hold-P	Cool-T	Dim. Sh. (mm)	S/N
	A	B	C	D	E	sh	Ratio
1	50	205	60	70	20	1.4	-2.9
2	50	210	65	80	25	1.5	-3.5
3	50	215	70	90	30	1.7	-4.6
4	50	220	75	100	35	1.9	-5.5
5	55	205	65	90	35	1.5	-3.5
6	55	210	60	100	30	1.6	-4.0
7	55	215	75	70	25	2.0	-6.0
8	55	220	70	80	20	2.1	-6.4
9	60	205	70	100	25	1.7	-4.6
10	60	210	75	90	20	1.9	-5.5
11	60	215	60	80	35	2.1	-6.4
12	60	220	65	70	30	2.3	-7.2
13	65	205	75	80	30	1.8	-5.1
14	65	210	70	70	35	2.0	-6.0
15	65	215	65	100	20	2.2	-6.8
16	65	220	60	90	25	2.3	-7.2

In order to know the significance of process parameters on dimensional shrinkage on product, ANOVA analysis [10] was performed and it was observed that melt temperature was most significant factor and cooling time is least significant factor. The rank is assigned to the factors based on delta statistics. The value with highest delta is ranked highest and lowest delta with least rank.

Table 3: Response table for DS (Raw data)

Level	A	B	C	D	E
1	-4.157	-4.040	-5.171	-5.550	-5.448
2	-5.017	-4.800	-5.282	-5.379	-5.346
3	-5.966	-5.981	-5.421	-5.235	-5.258
4	-6.302	-6.622	-5.569	-5.279	-5.390
Delta	2.145	2.582	0.398	0.314	0.190
Rank	2	1	3	4	5

In figure 2 it is clear that factors which are ranked higher are having peak profile and factors which are having lower ranks are having low profiles. Our objective is to always have higher value of S/N ratio which can be achieved by the peak point of the factors in figure 2 which will give the optimal case for DS response as shown in table 4.

Table 4: Optimal case for DS response

A	B	C	D	E
50	205	60	90	30

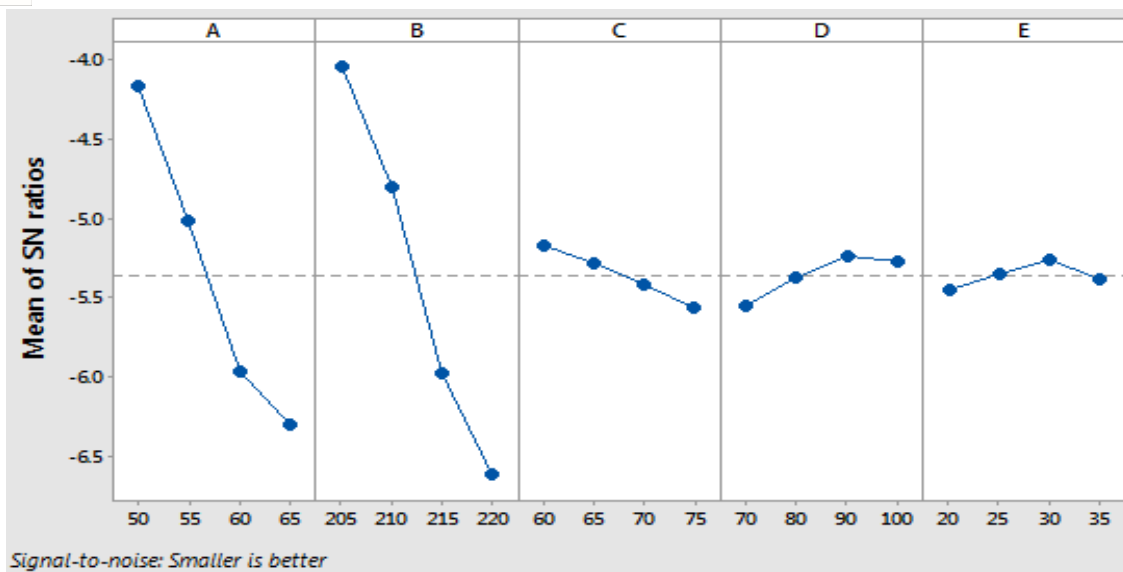


Figure 2: S/N ratio analysis for DS

Table 5: Analysis of variance for dimensional shrinkage

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	5	1.22150	0.244300	85.72	0.000
A	1	0.48050	0.480500	168.60	0.000
B	1	0.72200	0.722000	253.33	0.000
C	1	0.00450	0.004500	1.58	0.237
D	1	0.01250	0.012500	4.39	0.063
E	1	0.00200	0.002000	0.70	0.422
Error	10	0.02850	0.002850		
Total	15	1.25000			

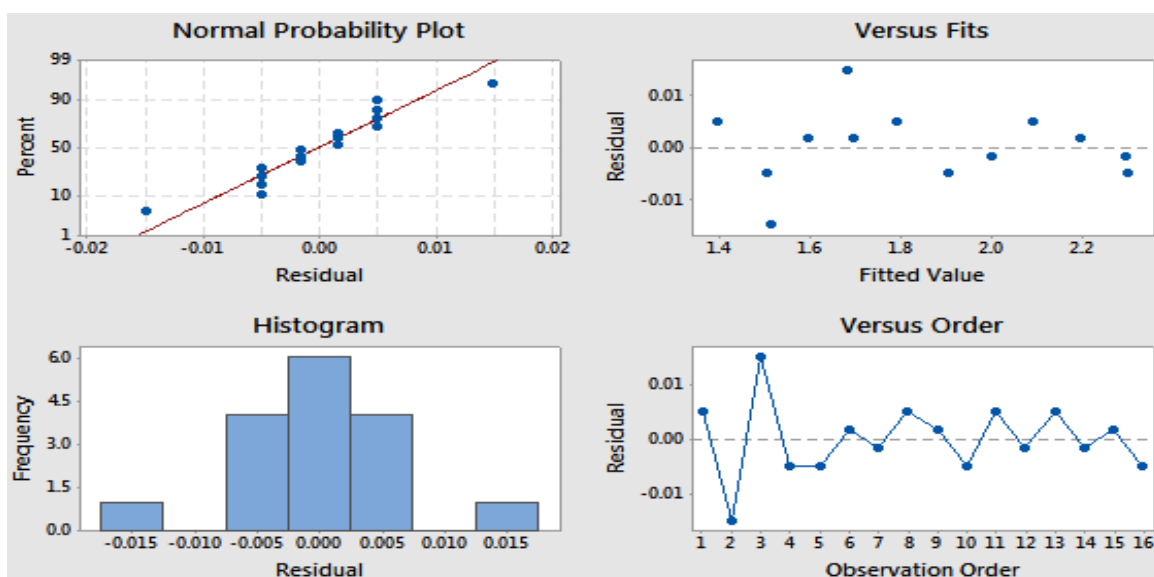


Figure 3: Residual Plots for DS

Figure 3 shows the residual plots for dimensional shrinkage. In normal probability plot, residuals follow the straight line. Histogram is also symmetric in nature which indicates normal distribution of residuals. In residual versus fitted value, residuals are scattered randomly around zero which shows that they possess constant variance. There is no error due to time or data collection order as no clear pattern is exhibited by residuals.

A. Optimum Value of Dimensional Shrinkage

The optimum value of process parameters have been found as A1, B1, C1, D3, E3 to minimize the dimensional shrinkage which is given by equation —

$$\mu_{ds} = A1 + B1 + C1 + D3 + E3 - 4T$$

Where, T = overall mean of dimensional shrinkage = $(\sum Ri) / 16 = 1.875$ mm

Where, values of R1 to R16 has been taken from the Table 2, and the values of A1, B1, C1, D3 & E3 has been taken from Taguchi experimental data.

A1 = average value of dimensional shrinkage at first level of mold temperature = 1.625 mm

B1 = average value of dimensional shrinkage at first level of melt temperature = 1.6 mm

C1 = average value of dimensional shrinkage at first level of Injection pressure = 1.85 mm

D3 = average value of dimensional shrinkage at third level of holding pressure = 1.85 mm

E3 = average value of dimensional shrinkage at third level of cooling time = 1.85 mm

Substituting the values of various terms in the above equation,

$$\mu_{ds} = 1.625 + 1.6 + 1.85 + 1.85 + 1.85 - 4 * 1.875 = 1.275$$

The 95 % confidence intervals of confirmation experiments (CI_{CE}) and population (CI_{POP}) are calculated as below:

$$CI_{CE} = \sqrt{F_{\alpha}(1, f_e) V_e \left[\frac{1}{n_{eff}} + \frac{1}{R} \right]} \quad \text{and}$$

$$CI_{POP} = \sqrt{\frac{F_{\alpha}(1, f_e) V_e}{n_{eff}}}$$

Where, $F_{\alpha}(1, f_e)$ = The F ratio at the confidence level of $(1-\alpha)$ against DOF 1 and error degree of freedom f_e .

$$n_{eff} = \frac{N}{1 + [\text{DOF associated in the estimate of mean response}]}$$

$$= 16 / (1+3) = 4$$

N = Total number of results = 16,

R = Sample size for confirmation experiments = 1

V_e = Error variance = 0.0028; f_e = error DOF = 10

(Table 5)

$F_{0.05}(1, 10) = 4.97$ (Tabulated F value ; Roy,1990)

$$CI_{CE} = \pm 0.1318, \text{ and } CI_{POP} = \pm 0.059$$

(Tabulated

Therefore, the predicted confidence interval for confirmation experiments is:

$$\text{Mean } \mu_{ds} - CI_{CE} < \mu_{ds} < \text{Mean } \mu_{ds} + CI_{CE}$$

$$\text{i.e. } 1.1432 < \mu_{ds} < 1.14068$$

The 95% confidence interval of the population is:

$$\text{Mean } \mu_{ds} - CI_{POP} < \mu_{ds} < \text{Mean } \mu_{ds} + CI_{POP}$$

$$\text{i.e. } 1.1216 < \mu_{ds} < 1.334$$

B. Confirmation Test

Confirmation test is very important in research which is designed by taguchi method. The objective of confirmation experiment is to know that how close theoretical value which is suggested by matrix experiment is to real data.

Table 6: Predicted optimal value and result of confirmation experiment

Response Parameter	Optimal Set of Parameters	Predicted Optimal Value	Predicted Confidence Intervals at 95% Confidence Level	Experimental Value
Dimensional Shrinkage	A ₁ B ₁ C ₁ D ₃ E ₃	1.275	CI _{POP} : 1.1216 < μ _{ds} < 1.334 CI _{CE} : 1.1432 < μ _{ds} < 1.4068	1.2

As the result is tabulated in table 6. It is clear that optimal value predicted by matrix experiment using taguchi method is very close to experimental value.

IV. CONCLUSION

Influence of plastic injection molding machine process parameters on dimensional shrinkage was studied in this research. The process parameters included in this study are mold temperature, melt temperature, injection pressure, holding pressure and cooling time. Significant variables affecting the dimensional shrinkage were identified by ANOVA analysis. Result shows that most significant variable affecting dimensional shrinkage is melt temperature and least significant variable is cooling time. A₁ = 50, B₁ = 205, C₁ = 60, D₃ = 90 and E₃ = 30 is the optimized value of machine process parameters for the present study.

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