# Processing parameters Optimization of Injection Moulding in DN20 Vent of Water Meter Manufacturing

N. Sateesh<sup>1\*</sup>, S. Devakar Reddy<sup>2</sup>, Ram Subbiah<sup>3</sup>, D. Siva Nagaraju<sup>4</sup>, BCh Nookaraju<sup>5</sup>

<sup>1</sup>Professor, Department of Mechanical Engineering, Gokaraju Rangaraju Institute of Engineering and Technology, Hyderabad,, India <sup>2</sup>M.Tech, Engineer, Tech Mahindra LTD., Hyderabad India

<sup>3,4,5</sup>Associate Professor, Department of Mechanical Engineering, Gokaraju Rangaraju Institute of Engineering and Technology, Hyderabad,, India

**Abstract.** The conventional optimization process in Injection Moulding includes actual shop floor trials in which melt temperature, mould temperature, injection time, injection pressure, pattern, feeder size, shape and location cores, mould layout, gating etc. are changed in each iteration which involves high machining cost, tooling cost, modification cost, melting cost, and transportation cost as well as, materials, energy, time are wasted in each trial until and unless the required results are obtained. Water meter component (DN20 Vent) is designed in CREO 5.0, and then components are 3D printed to cross check the dimensions and also to confirm whether all the other components can be accommodated or not. Then the mould flow analysis will be performed on a water meter components using different materials and changing the processing parameters. The input processing parameters considered are melt temperature, mould temperature and injection time, whereas the responses are warpage, volumetric shrinkage, cycle time and quality prediction. Grey relational analysis is carried out to determine the optimum injection moulding processing parameters. The effort has been made to minimize the warpage, volumetric shrinkage, cycle time and maximize the quality prediction mould cavity and core for the components are designed in CREO 5.0 and manufactured using P20 tool steel. Then the water meter components are manufactured by inputting the optimal processing parameters in injection moulding machine to achieve high productivity and quality.

# **1** Introduction

Products of highest quality can be obtained with enhance mould design by using Mould Flow Analysis (MFA) software which replicates flow of the plastic inside the mould cavity. Potential areas of concern can be highlighted by this analysis as it offers results of how the selected material fills the mould cavities. An analysis of the mould flow indicates potential problems associated with moulding and can be corrected before cutting steel to make the mould so that expensive and laborious tooling rework can be prevented.

In beginning, mould is designed and evaluate to ensure uniform parts production from cavity using MFA. Flow of Resin in mould cavity is forecasted by model developed and resin characteristics. Mould processing parameters like melt temperature, pressure profile, or filling time, injection pressure are optimized before the mould is manufactured. Optimum processing parameters, shorter cycle times, shorter filling times and fewer defects will be analysed. This optimum parameters are used in the manufacture of the DN20 vent. Vishwas Lomate, Salunke M, K. Rushikesh, S.Gajanan 2015, [1] have performed modelling of mould flow on the plastic part with deviations such as shrinkage, weld lines, air traps and immersion marks in manufacture of toy. Sanusi Md, Aziz, Ali Amran, Idayu N, Hadzley Md, and S Sivarao 2016 [2] has carried out simulation and found optimized injection mould melt temperature. S.Rajalingam, Awang Bono and Jumat bin Sulaiman 2013, [3] optimized processing parameters such as (screw rotation speed, injection pressure and mould temperature in manufacture of mobile phone case and investigated the affect the shrinkage defect of the plastic case.

MD Helen, Huszar M, Belblidia F, Arnold C, David Bould, Johann Sienz 2018, [4] investigated and suggested common defect i.e warpage in injection moulding process. Gurjeet, Pradhan M K, Ajay Verma 2018 [5] proposed an approach for multifactor optimization of parameters of the injection moulding process such as packing time, injection pressure, cycle time and melting temperature. Satyanarayana Kosaraju, Vijay Kumar M, Sateesh N. 2016, [6] employed multiobjective optimization based on Taguchi-based Grey relational method, to find the optimal levels of cutting

Corresponding author: <u>sateeshnagari1@gmail.com</u>

<sup>©</sup> The Authors, published by EDP Sciences. This is an open access article distributed under the terms of the Creative Commons Attribution License 4.0 (http://creativecommons.org/licenses/by/4.0/).

parameters for the objective of lower cutting force and better surface finish under dry cutting conditions for Inconel 625. Tosun 2006, [7] determined of optimum parameters for multi-performance characteristics in drilling by using grey relational analysis. A.Noorul Haq, P.Marimuthu, R.Jeyapaul, 2008, [8] presented a new approach for the optimization of drilling parameters on drilling Al/SiC metal matrix composite with multiple responses based on orthogonal array with grey relational analysis. Experiments are conducted on LM25-based aluminium alloy reinforced with green bonded silicon carbide of size 25 µm (10% volume fraction). Drilling tests are carried out using TiN coated HSS twist drills of 10 mm diameter under dry condition. V.Balaji, N.Sateesh, M. Manzoor Hussain, 2015, [9] manufactured of Aluminium Metal Matrix Composite (Al7075-SiC) by Stir Casting Technique. N.Sateesh, P.Sampath Rao, D.V.Ravishanker, K.Satyanarayana, 2015. [10] conducted to investigate the environmental impacts of fiber composites. The main objective of this work is to investigate the degradation of GFRP composite which is exposed to different environmental conditions and its influence on the tensile strength. Ratna Deepika Manikonda, Satyanarayana kosaraju, K. Arul Raj, N.Sateesh 2018, [11] investigated Wear Behavior Analysis of Silica Carbide Based Aluminum Metal Matrix Composites using pin on disc.

# 2. Experimentation

### 2.1 DN20 Vent Component Modelling

DN20 vent component is designed in CREO 5.0 as shown in fig.2.1. Processing parameters levels considered in mould flow analysis is shown in table 2.1.

Table.2.1. Processing parameters levels for DN20 vent

	LEVELS		
PARAMETER	1	2	3
Melt Temperature( <sup>0</sup> C)	280	300	320
Mould Temperature( <sup>0</sup> C)	70	80	90
Injection Time(Sec)	0.4	0.7	1.0

# 2.2 Mould flow Analysis (MFA) of DN20 vent

MFA of DN20 vent is performed using mould flow adviser 2016 software. The material is used is Nylon 6 and its properties are considered for analysis. Three levels of each processing parameters are considered as melt temperature (280°C, 300°C, and 320°C), mould temperature (70°C, 80°C, 90°C) and injection time (0.4s, 0.7s, 1.0s).

When solidification begin, cooling state to ejection state part volume is reduced which is known as volumetric shrinkage. The sum of mould opening time, cooling time fill and pack time is known as cycle time. Like that MFA is done for nine different inputs of processing parameters as shown in table 2.2 and the corresponding results of MFA shown in same table 2.2. The results like deflection, volumetric shrinkage and deflection (warpage) shown in the table 2.2.



**Fig.2.1**. Result of analysis of DN20 vent at melting temperature = $280^{\circ}$ C, mould temperature =  $70^{\circ}$ C, injection time = 0.4 s.

Table.2.2.	MFA	Results	of DN20	vent

S N 0	Mel t Tem pera ture (°C)	Mo uld Tem pera ture ( <sup>0</sup> C)	Inje ctio n Tim e (Sec )	Deflec tion (mm)	Volume tric Shrinka ge at Ejectio n (%)	Cycle Time (Sec)	Qua lity Pre dicti on (%)
1	280	70	0.4	1.425	13.24	17.65	46.1
2	280	80	0.7	1.426	13.35	18.20	16.6
3	280	90	1.0	1.438	13.47	18.76	7.55
4	300	70	0.7	1.357	13.12	15.72	95.7
5	300	80	1.0	1.416	13.33	16.03	95.4
6	300	90	0.4	1.390	13.14	15.41	95.4
7	320	70	1.0	1.463	14.72	20.26	78.6
8	320	80	0.4	1.464	14.83	20.9	78.1
9	320	90	0.7	1.469	14.88	21.71	78.2

# 2.3 Mould Design and manufacturing of DN20 Vent

Core and cavity of DN20 vent is modelled using CREO software shown in fig.2.1, 2.2 2.3, 2.4 and 2.5.





Fig.2.2. Core

Fig.2.3. Cavity sub insert



Fig.2.4. Core and cavity sub insert of DN20 vent



Fig.2.5. DN20 Vent core

# 3. Grey Relational-Analysis

# 3.1 Dn20 Vent manufacturing results:

Considering table 2.2 results and using the grey relational analysis method, the normalized values of cycle time, volumetric shrinkage, quality prediction and deflection can be found. The same is shown in table 3.1.

The qualitative characteristic "smaller the better". Hence it is used to minimize the cycle time, volumetric shrinkage, and deflection. "larger is better" is used to maximize the quality prediction. The equations are used for finding normalized values are taken from grey relational analysis method. Table 3.1 presented the results and table 3.2 shows deviation sequences.

**Table.** 3.1. DN20 vent- cycle time, volumetricshrinkage, quality prediction and deflection

S. No	Deflection (mm)	Volumetric Shrinkage (at Ejection) (%)	Cycle Time (Sec)	Quality Predicti on (%)
1	0.3928	0.9318	0.6444	0.4373
2	0.3839	0.8693	0.5571	0.1026
3	0.2767	0.8011	0.4682	0
4	1	1	0.9508	1
5	0.4732	0.8806	0.9016	0.9965
6	0.7053	0.9886	1	0.9965
7	0.0535	0.0909	0.2302	0.8060
8	0.0446	0.0284	0.1285	0.8003
9	0	0	0	0.8015

<b>Table 3.2</b> I	DN20 vent	- Deviation	sequences
--------------------	-----------	-------------	-----------

S.	Δ0i (1)	Δ0i (2)	$\Delta 0i(3)$	$\Delta 0i(4)$
No.				
1	0.6072	0.0682	0.3556	0.5627
2	0.6161	0.1307	0.4429	0.8974
3	0.7233	0.1989	0.5318	1
4	0	0	0.0492	0
5	0.5268	0.1194	0.0984	0.0035
6	0.2947	0.0114	0	0.0035
7	0.9465	0.9091	0.7698	0.194
8	0.9554	0.9716	0.8715	0.1997
9	1	1	1	0.1985

Since all links are given the same choice, it is considered  $\xi$  as 0.25. Coefficient of gray relativity for each experiment is found and shown in table 3.3. Based on highest grey relational grade (i.e Rank 1) experiment no

4 has the best multiple performance characteristics among all nine experiments. In present study optimum processing parameters of injection moulding can be found from grey relational grade.

Table 3.3 Grey relational-grade values and order

	Gr	Grey	R			
	Deflecti	Volume	Cycle	Quality	relation	a
S.	on	tric	Time (s)	Predicti	al grade	n
Ν	(mm)	Shrinka		on (%)	(yi)	k
0		ge at				
		Ejectio				
		n (%)				
1	0.2916	0.7856	0.4128	0.3076	0.4494	4
2	0.2886	0.6566	0.3608	0.2178	0.3809	5
3	0.2568	0.5569	0.3197	0.2	0.3335	6
4	1	1	0.8355	1	0.9588	1
5	0.3218	0.6768	0.7175	0.9862	0.6755	3
6	0.4589	0.9564	1	0.9862	0.8504	2
7	0.2089	0.2157	0.2451	0.5631	0.3802	7
8	0.2074	0.2046	0.229	0.5559	0.2992	8
9	0.2	0.2	0.2	0.5574	0.2893	9

It can be distinguished the influence of each processing parameter on the gray relational bias at different levels since experimental design is orthogonal. By averaging the relative gray estimate for experiments 1, 2, and 3, 4, 5 and 6, 7, 8 and 9, respectively, average value of the relative gray estimate for the melt temperature at levels 1, 2, and 3 will be found. The same is shown in Table 3.4. Response graph (Signal-to-noise ratio) of gray relational estimate shown in Figure 3.1. The optimal level of processing parameters is the level with the highest degree of relativity of gray.

The average value of the relative degree of greyness for each level of processing parameters and total average gray value for nine experiments are presented in table 3.4. Optimized processing parameters (less volume shrinkage, cycle time, deflection, and high-quality prediction) correspond to [A2B1C2], is shown in table 3.4.

Table 3.4 Grey relational grade -Response table

S y m b o l	Proce ssing para meter s	Grey 1 LEVEL 1	Grey Relational Grade LEVEL LEVEL LEVEL 1 2 3		Main Effect (max – min)	R a n k	
	Melt	0.3879	0.8282*	0.2989	0.5293	1	
А	Tempe						
	rature						
	Mould	0.5721*	0.4518	0.4911	0.1203	2	
В	Tempe						
	rature						
	Injecti	0.533	0.543*	0.4391	0.1034	3	
С	on						
	Time						
	Total ave	rage grey re	lational gra	de values (γ	m = 0.513)		
	*	grey relation	onal grade-o	ptimum valı	ues		



Fig. 3.1. S/N ratios response graph of Grey Relational Grade

#### 3.2 Analysis of Variance (ANOVA)

ANOVA is being carried out to investigate which factor is most influence on performance. It is achieved by isolating the relational estimates of gray total variability. This is calculated by the sum of the squares of the deviations from the total average value of the relational estimates of gray for each process parameter and error. The contribution % of each factor to sum of square of deviations can be used to assess the processing parameters performance. The F value shown in Table 3.5 will also be used to find the factor which has a more influence on performance. A change in a certain factor has more influence on the performance profile, when the F value is large.

Table 3.5Processing parameters - contributionpercentage

Source	DF	Adj SS	Adj MS	F- Valu e	Percen tage contri bution (%)
Melt Temper ature	2	0.453455	0.226728	57.67	89.6
Mould Temper ature	2	0.033389	0.016694	4.25	6.60
Injectio n Time	2	0.011380	0.005690	1.45	2.25
Error	2	0.007863	0.003931		1.55
Total	8	0.506086			100

It is obvious that melt temperature is substantial parameter and is given in table 3.5. Table 3.3 shows that 0.09588 was the highest score of grey scale from experiment run 4.

Average grey relational grade from Mould flow analysis results shown in table 2.2. it can be observed that the highest grey relational grade value i.e mould temperature, melt temperature, and injection time are 70°C, 300°C, and 0.7s, respectively.

In order to achieve best quality, lower volumetric shrinkage, less deviation, and lesser cycle time these are few levels of controlled technological factors that are recommended. This analysis shows that most influence processing parameter is melting temperature, next by injection time, and mould temperature which affect the injection moulding of DN20 Vent.

#### 3.3 DN20 Vent- Injection Moulding



Fig.3.2. Dn20 vent mould core and cavity manufacturing process.



(melting temperature is 300°C, mould temperature is 70°C, injection time is 0.7s.)

DN20 vent mould cavity, and core are manufactured by injection moulding process shown in fig. 3.2. DN20 vent is manufactured at optimized processing parameters where mould temperature is 70°C, melting temperature is 300°C, and injection time is 0.7 s as presented in fig. 3.3.

#### 4. Conclusion

Water meter components of DN20 Vent were developed in CREO software. The components are printed in 3D to evaluate the dimensions and to cross check whether all components can be placed in mould or not. MFA is carried out on the components of the water meter using nylon 6, 6 taking into account 3 levels of each processing parameter, i.e., the melting point (280°C, 300°C, 320°C), mould temperature (700°C, 800°C, 900°C) and injection time (0.4s, 0.7s, 1.0s) using Mould Flow Adviser sofware. From the results of MFA, it was found that the responses are cycle time, deflection, shrinkage volume, and quality prediction. According to the grey relational analysis results, the highest degree of gray for DN20 Vent, processing parameters are mould temperature 70°C, melting temperature 300°C, and injection time 0.7s. The mould cavity and core components are developed in CREO and made of P20 tool steel. DN20 Vent components are manufactured

with optimized processing parameters by injection moulding machine.

Today: Proceedings, Volume 5, Issue 9, Part 3, Pages 20104-20109

# **References:**

- 1. Vishwas Lomate, Salunke M, K. Rushikesh, S.Gajanan "Injection Moulding Methods Design, Optimization, Simulation of Plastic Toy Building Block by Mould Flow Analysis". International Journal of Mechanical Engineering and Technology (2015).
- Sanusi Md, Aziz, Ali Amran, Idayu N, Hadzley Md, and S Sivarao. "Optimization of Processing parameters in Linear Runner Family Injection Mould Using Mould Flow Simulation Software". ARPN Journal of Engineering and Applied Sciences (2016)
- S.Rajalingam, Awang Bono and Jumat bin Sulaiman "A statistical experimental study on shrinkage of injection-moulded part" International Journal of Humanities and Management Sciences (IJHMS) Volume 1, Issue 1, 2013 pp 2320–4044
- MD Helen, Huszar M, Belblidia F, Arnold C, David Bould, Johann Sienz. "Sustainable Injection Moulding: The Impact of Materials Selection and Gate Location on Part Warpage and Injection Pressure". Sustainable Materials and Technologies 5 (2015) 1–8.
- 5. Gurjeet, Pradhan M K, Ajay Verma "Multi Response optimization of injection moulding Processing parameters to reduce cycle time and warpage", Materials today: proceedings · January 2018.
- Satyanarayana Kosaraju, Vijay Kumar M, Sateesh N. "Optimization of Machining Parameter in Turning Inconel 625", 6th International Conference of Materials Processing and Characterization (ICMPC 2016).
- N. Tosun, "Determination of optimum parameters for multiperformance characteristics in drilling by using grey relational analysis", International Journal Advanced Manufacturing Technology, 28, 2006, pp. 450-455.
- 8. A.Noorul Haq, P.Marimuthu, R.Jeyapaul, "Multiresponse optimization of machining parameters of drilling Al/SiC MMC using grey relational analysis in the Taguchi method, 2008, pp. 250-255.
- V.Balaji, N.Sateesh, M. Manzoor Hussain, Manufacture of Aluminium Metal Matrix Composite (Al7075-SiC) by Stir Casting Technique, 2015, Materials Today Proceedings, Volume 2, Issues 4–5, pp 3403-3408.
- N.Sateesh, P.Sampath Rao, D.V.Ravishanker, K.Satyanarayana, 2015, Effect of Moisture on GFRP Composite Materials, Materials Today Proceedings, Volume 2, Issues 4–5, pp 2902-2908
- 11. Ratna Deepika Manikonda, Satyanarayana kosaraju, K. Arul Raj, N.Sateesh, (2018), Wear Behavior Analysis of Silica Carbide Based Aluminum Metal Matrix Composites, Materials