

Optimization of Thinning In Deep Drawing Process Using Ant-Lion Optimization Algorithm (May 2016)

Anuja S.Joshi ¹, M.E.Student , Prof A.U.Gandigude ²,Asso. Prof. Mechanical Department (ZCOER)

Abstract —Deep drawing is a manufacturing process which is used extensively in the forming of sheet metal into cup as well as box like structures. In deep drawing process the depth of the part being made is more than half its diameter. The geometry of die which influences the thickness distribution and thinning of sheet metal blank in the deep drawing processes. Excess thinning in deep drawing process is caused by incorrect die and punch clearance and radii. Thinning will usually be greatest near the base of part. Thinning effect created by tension forces. Determination of the thickness distribution and of the thinning of the sheet metal blank results into reduction of production cost of the material and time too. In general the final objective of deep drawing process in particular or of any sheet metal forming process is to produce good quality product, hence uniform thickness should be obtained throughout. Ant Lion optimization algorithm which is used here which is developed by Seyedali Mirjalili in Jan.2015. Ant Lion optimization algorithm is used to obtain optimal effect of blank holding force on the thinning of deep drawing parts.

Keywords: Deep Drawing Process , Thinning, Ant lion Optimization

I. NOMENCLATURE

BHF Blank holding force(N)
 F_{dmax} Forming load(N)
 R_D Radius on Die(mm)
 R_p Radius on Punch(mm)
 μ Coefficient of Friction

 d_1 and d_2 Diameters of component at bottom and at the top respectively (mm)
 d_0 Blank diameter (mm)
 r Corner radius (mm)
 S_0 Material thickness (mm)
 S_u Ultimate tensile strength (495 N/mm²)
 P Pressure applied (2.5 N/mm²)

II. INTRODUCTION

Deep drawing is a mostly used metal forming process in which a sheet metal blank is drawn into a forming die by the mechanical action of a punch. Deep drawing has been classified into conventional and unconventional deep drawing. The sheet metal work piece i.e. blank is placed over the die

opening. A blank holder applies pressure to the entire surface of the blank except the area which comes under the punch, holding the sheet metal work flat against the die [1, 2]. The punch travels towards the blank. After contacting the blank, the punch forces the sheet metal into the die cavity, forming its shape. It is thus a shape transformation process with material retention. The process is considered as "deep" drawing process when the depth of the drawn part exceeds its diameter.

Deep drawing process is capable of forming circular shapes such as circular cups or even rectangular shapes or shell-like containers. The deep drawing process which occurs under a combination of both tensile and compressive forces is a forming process. Wall thinning is prominent due to tensile forces and results in an uneven part wall thickness. The material starts to flow in the die when the stress exceeds. Determination of the thickness distribution of the sheet metal blank reduces the production cost of the material and time. The final objective of deep drawing process in particular or of any sheet metal forming process in general is to produce good quality product, hence uniform thickness should be obtained throughout.

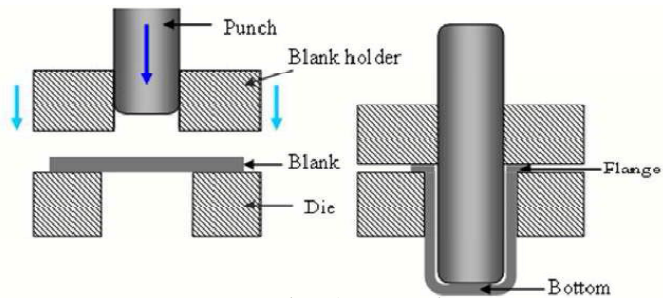


Figure 1. Deep Drawing Process

III. LITERATURE REVIEW

H. Zein, M. El Sherbiny, M. Abd-Rabou, M. El Shazly [4] studied on thinning and spring back prediction. Prediction of the forming results as spring back, determination of the thickness distribution and of the thinning of the sheet metal blank reduces the production cost of the material and time.

Y. Marumo, H. Saiki, L. Ruan [5] studied the influence of sheet thickness on blank holding force and limiting drawing ratio. Variation in blank holding force and limiting drawing ratio in deep drawing of metal foils were evaluated by calculation. The paper shows variation in the blank holding force required for the elimination of wrinkling and the limiting drawing ratio with sheet thickness. Menakshi Mahendru

Anuja S.Joshi ¹ is with the M.E. student, ZCOER, Pune-411041 (e-mail: anuja.joshi30@gmail.com).

Nischal,Shivani Mehta[11]solved optimal load dispatch problemby using ALO. Petrovi, M., Petronijevi, J.etc[12] used ALO for flexible process planning. S. Talatahari[13]worked on optimal design of skeletal structure to design three truss and frame design optimization. Shivani Mehta, Meenakshi Mahendru Nischal[14]used ALO for optimumpower generation so that operation of power is most efficient and at low cost. Seyedali Mirjalili[15] had done very much work on development of Ant Lion Optimizer, its behavior, pseudo code and solved constrained optimization problemby applying ALO.

IV.TH INNING IN DEEP DRAWING

The original blank thickness has some effect on the thinning and thickness distribution of sheet metal blank in the deep drawing processes. The average distribution of the wall thickness is increasing with increasing the blank thickness. As blank thickness increases % of thinning also increases with respect to it. Taking into account, the punch diameter and blank thickness effects the limiting drawing ratio (LDR) decreases as the relative punch diameter increases. During the deep drawing process slightly thicker materials can be gripped in better way. With increasing in thinning sheets stretched to a greater extend because of its more volume.[3]

A. Blank Holder Force (BHF)

Blank holder force (BHF) is a very important parameter in thedeepdrawing process. It is used to suppress the formation of wrinkles which appear on the flange of the drawn part. Also when we increases the BHF, stress normal to the thickness increases which restrains any formation of wrinkles. In order to have less thinning in the drawn part, the maximumpunch force must be reduced and this can be achieved by controlling the value of the BHF throughout the process.[4]

B. Radius on Die (R_D)

Theoretically, the radius on the draw die also called as draw ring should be as large as possible to permit full freedom of metal flow as it passes over the entire radius. Draw ring causes the metal to begin flowing plastically and side near compressing and thickening of outer portion of the blank. However, if the draw radius is too large, the metal will be release through the blank holder too soon and results into wrinkling . Too sharp radius will hinder the normal flow of the metal and causes uneven thinning of a cup wall, with resultant erring.[5]

C. Radius on Punch (R_P)

There is no such a set rule for the size of the radius on the punch. Sharper radius will require higher forces when the metal is folded around the nose of punch and may result in excessive thinning and some times tearing at the bottom of the cup. A general rule to make reduction in the thinning is to design the punch with a radius offrom 4-10 times the metal thickness.

D. Coefficient of Friction (μ)

The force of static friction between the work piece, blank and draw die surface must overcome in a drawing operation

and the force of the blank holder adds significantly to the force of static friction.[6]

V.CO MPONENT DESCRIPTION

The component selected for thinning optimization is sealing cover. The thickness of sheet is 2 mm.

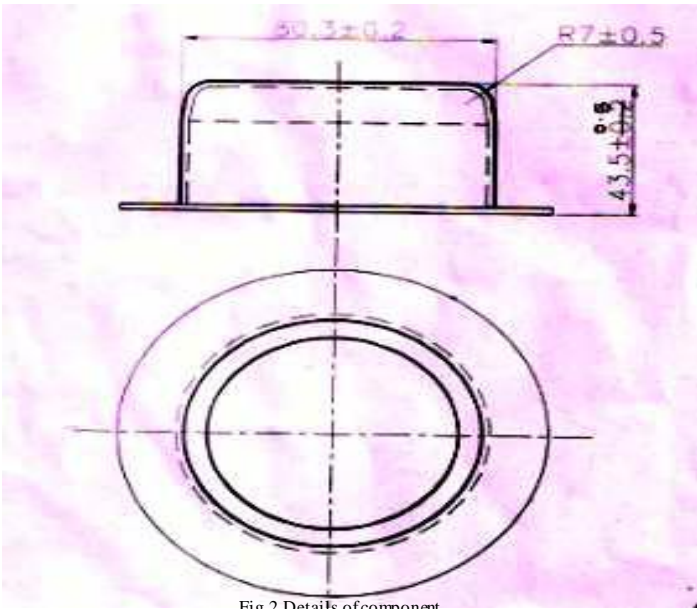


Fig.2 Details of component

VI. PROBLEM FORMULATION

From component description diagram, we get values of d₁,d₂,d₃,r,h.Using these values equations explained below are solved.

Forming load,
$$F_{\text{max}} = 1.2 (d_1 + S_0) \cdot S_0 \cdot S_u \tag{1}$$

Blank holder force,
$$FBH = (d_0^2 - (d_1 + 2r)^2) \cdot P \tag{2}$$

Blank diameter, d₀
$$d_0 = \sqrt{(d_1 + 2r)^2 + \frac{FBH}{P}} \tag{3}$$

Radius of Die,
$$R_D = 0.035 [50 + (d_0 - d_1)] \tag{4}$$

Radius of Punch,
$$R_P = 3 \text{ to } 6 R_D \tag{5}$$

The above values are calculated by numerically and by using software (Forming Suit) also, the values came by these both methods are nearby same. Values are tabulated below.

TABLE NO.1 EXPERIMENTAL RESULTS				
Exp.no.	Blank holder force	Coe.of friction, μ	Radius of Punch,R _P	
1	600 0.05 7			
2	600 0.10 8			
3	600 0.15 9			
4	700 0.05 8			
5	700 0.10 9			
6	700 0.15 7			
7	800 0.05 9			

$$\begin{matrix} 8 & 800 & 0.10 & 7 \\ 9 & 800 & 0.15 & 8 \end{matrix}$$

By using above exp. results we had done regression analysis of above values, then we get equation for thinning. The optimization problem for thinning in deep drawing is formulated as follows:
Thinning= $0.113+0.000314BHF-0.0366 \mu-0.00016R_D+0.00025r$
Subjected to:
 $4.2 < R_D < 8.5$
 $3 R_D > R_p > 6 R_D$
The problem was solved using by Ant-Lion optimization algorithm and results were obtained.

VII.ANT LION OPTIMIZATION ALGORITHM

The names of ant lions originate from their unique hunting behavior and their favorite prey. Ant lion larvae dig a cone-shaped pit in to sand by moving along a circular path and throwing out sands outside with its massive jaw [11,12]. After trap digging, the larvae hides underneath the bottom of the cone and waits for insects (preferably ant) to come to be trapped in to the pit [13]. The edge of the cone is sharp so insects fall to the bottom of the trap easily. When the ant lion realizes, a prey is in the trap, it tries to catch it with its massive jaw. When a prey is caught into the jaw, it is pulled under the soil easily and consumed. After consuming the prey, ant lions throw the leftovers outside the pit and dig the pit for the next hunt. Also another interesting behavior that has been observed in life style of ant lions is that they tend to dig out larger traps as they become more hungry and/or when the moon is full [9]. They have been evolved and adapted by this way to improve their chances of survival. The main inspiration of the ALO algorithm comes from the foraging behavior of ant lion's larvae. The ALO algorithm mimics interaction between ant lions and ants in the trap. To model such interactions systematically, ants are required to move over the search space, and ant lions are allowed to hunt them and become fitter using traps. [15]



Fig.3 Cone-shaped traps and hunting behaviour of ant lions

Since ants move randomly in nature when searching for food, a random walk is chosen for modeling ants' movement as follows:
 $X(ite r)=[0,cumsum(2r(1)1),cumsum(2r(2)1),...,cumsum(2r(n)-1)]$
Where *cumsum* calculates the cumulative sum, *n* is the max. no. of iteration, *ite r* shows the iteration of random walk, and *r(t)* is a stochastic function defined as follows:
 $r(t)=1 \text{ if } rand>0.5$
 $r(t)=0 \text{ if } rand \leq 0.5$

The position of ants and their related objective functions are saved in the matrices *MAnt* and *MOA*, respectively. In addition to ants, it is assumed that the ant lions are hidden somewhere in a search space, in order to save their positions and fitness values, the *MAntlion* and *MOAl* matrices are used. The pseudo codes the ALO algorithm is defined as, [1]:

Step 1: Initialize the first population of ants and ant lions randomly in search space and calculate the fitness of ants and ant lions.

Step 2: Find the best ant lions and assume it as the elite. In this study the best ant lion obtained so far in each iteration is saved and considered as an elite.

Step 3: For each ant, select an ant lion using Roulette wheel and

3.1 Create a random walk
3.2 Normalize this in order to keep the random walks inside the search space
 $X_i^{ite r+1}=[(X_i^{ite r}-a_i) \times (d_i-a_i)] \div (b_i-a_i)+c_i \dots(9)$
where *a_i* is the min. of random walk of *i*-th variable, *b_i* is the maximum of random walk in *i*-th variable, *c_i* and *d_i* are the min. and max. of *i*-th variable at the current iteration.

3.3 Update the position of ant
 $Ant_i^{ite r+1}=(R_A^{ite r}+R_E^{ite n}) \div 2 \dots(10)$

where *R_A^{ite r}* is the random walk around the ant lion selected by the roulette wheel; *R_E^{ite n}* is the random walk around the elite and *Ant_i^{ite r+1}* indicates the position of *i*-th ant at the iteration *ite r*+1.

3.4 Update *c* and *d* using the following
 $C^{ite r+1}=C^{ite n} \div I \dots(11)$

$d^{ite r+1}=d^{ite n} \div I \dots(12)$

Where
 $I=10^{w \cdot ite r/n} \dots(13)$

And *w* is a constant based on the current iteration (*w* = 2 when *ite r* > 0.1 *n*, *w* = 3 when *ite r* > 0.5 *n*, *w* = 4 when *ite r* > 0.75 *n*, *w* = 5 when *ite r* > 0.9 *n*, and *w* = 6 when *ite r* > 0.95 *n*)

Step 4: Calculate the fitness of all ants.

Step 5: Replace an ant lion with its corresponding ant if it becomes fitter.

Step 6: Update elite if an ant lion becomes fitter than the elite.

Step 7: Repeat procedure from step 3 until a stopping criteria is get satisfied

The pseudo code of the ALO algorithm-

- 1) Initialize the first population of ants and ant lions randomly
- 2) Calculate the fitness of ants and ant lions
- 3) Find the best ant lions and assume it as the elite (determined optimum)
- 4) **While** the end criterion is not satisfied
- 5) **For**
Every ant
 - (i) Select an ant lion using Roulette wheel
 - (ii) Update *c* and *d* (max & min. variable) using equations
 - (iii) Create a random walk and normalize it using equation.(9)
 - (iv) Update the position of ant
- 6) **End for**
 - (i) Calculate the fitness of all ants
 - (ii) Replace an ant lion with its corresponding ant if.....(8)

ant becomes fitter
(iii) Update elite if an ant lion becomes fitter than the elite
End while
Return elite

VIII .OPTIMIZATION RESULTS

The formulated optimization problem was solved by Ant Lion Optimization algorithm. The formability analysis was done on the original component and the forming limit diagrams were showing the results of thickness distribution, safety zone and forming zone of original component.

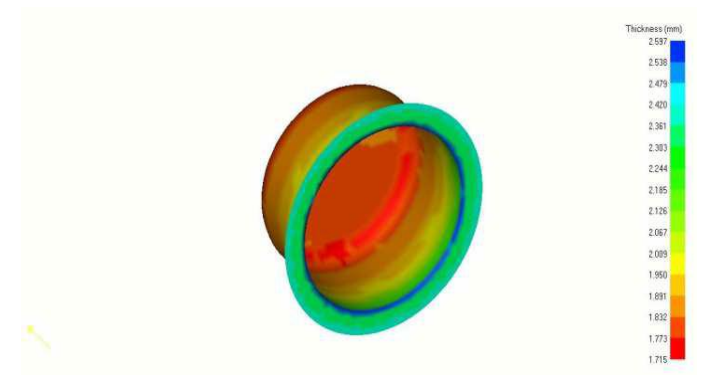


Figure 4. Thickness distribution of Original Component

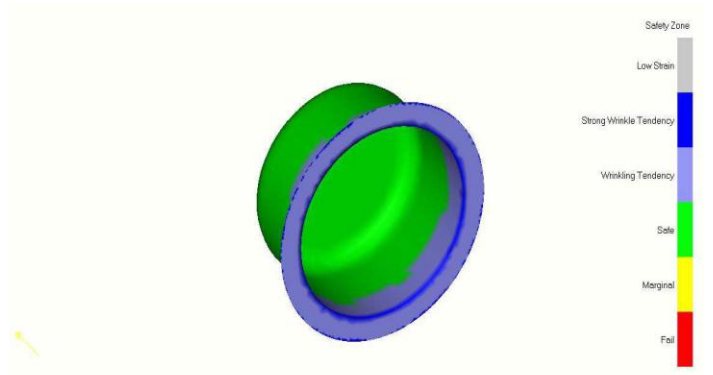


Figure 5. Safety Zone of Original Component

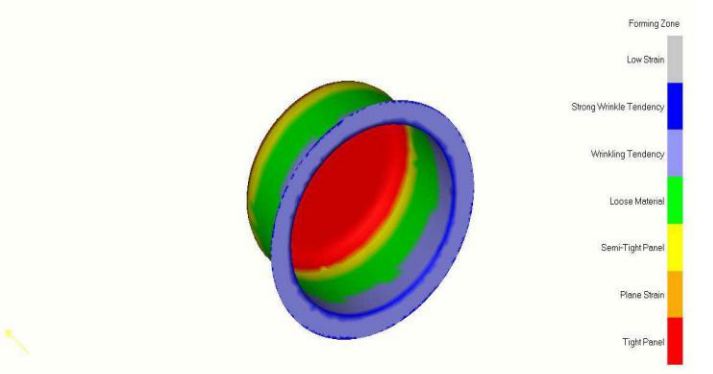


Figure 6. Forming Zone of Original Component



Figure 7. Principal curvature direction of Original component

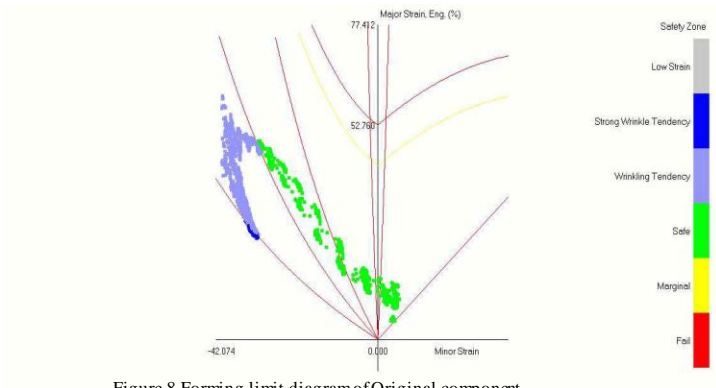


Figure 8. Forming limit diagram of Original component
The further work is ongoing on thinning optimization.

IX. CONCLUSIONS

Excessive thinning in areas of the sheet metal is mostly an unwanted defect. Maximum thinning will most likely occur at the side wall, near by the base of the part. The parameters affecting on thinning in deep drawing are blank holder force, radius on die, radius on punch and coefficient of friction. By controlling all these parameters minimization of thinning occurs. Here, in this paper Ant Lion optimization algorithm is used to optimize the thinning in deep drawing. Here four input parameters are blank holder force, radius on die, radius on punch and coefficient of friction. The results show the thickness distribution of original component, safety zone of original component and forming zone of original component. And from this analysis it comes to know that this algorithm gives optimized results of thinning.

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Miss. Anuja Joshi¹ received her Bachelor’s degree in Mechanical Engineering from Shivaji University and pursuing masters degree in CADME domain from SPPU, Pune. Her interests are co-inside into Optimization techniques.



Prof. A. U. Gandigude² is working as Asst. Prof & P.G teacher at ZCOER, Pune .He has been awarded for his research projects as a best guide. He had received many national and international awards at various national and international platforms .His areas of interests are Lean, Fluid power, Machine tools.