STOCHASTIC OPTIMIZATION OF MULTIPLEX CHARGES BASED ON INTERIOR BALLISTICS SIMULATIONS

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ABSTRACT

Multiplex charges combine propellant grains with regressive, neutral and progressive burning behaviors in a single propellant charge to improve the interior ballistics. This work optimizes the mass and the web thickness of each type of propellant grain used in a multiplex charge. Two optimization problems are studied: the maximization of the muzzle velocity and the minimization of the maximum pressure inside the gun. Particle Swarm Optimization (PSO) and Differential Evolution (DE) were used to solve the optimization problems. The interior ballistics of a 155 mm howitzer was simulated by using a lumped parameters model. In this model, the lost energy and the resistance pressure are defined by spline functions based on reference data. The results show that optimized multiplex charges can be used to increase the weapon's utility range and to reduce the weapon's weight.

Keywords: interior ballistics, multiplex charge, optimization, lumped parameter

NOMENCLATURE

- a pressure index of the propellant
- B burning rate constant of the propellant
- c co-volume
- CF DE mutation parameter
- CR DE crossover parameter
- d caliber
- D web thickness
- E energy
- f web fraction
- F propellant impetus
- IR inner radius
- k iteration
- m mass
- P gas pressure
- Q heat
- rand random number
 R constant of the gas
 V velocity of the projectile
- Vol volume
- S projectile position
- L barrel length
- t time
- T gas temperature
- z propellant form functionU internal energy of the gas
- W projectile kinetic energy
- x particle in the population
- y particle velocity in PSO

Greek symbols

- α PSO inertial parameter
- β_1 PSO self-learning parameter
- β_2 PSO global learning parameter
- γ gas specific heats ratio

- η thermal efficiency
- ρ propellant density

Subscripts

- atm atmospheric
- cc combustion chamber
- g gas
- i igniter
- j type of the propellant grain
- lost lost
- max maximum
- p propellant
- proj projectile r resistance
- ref 155M549 PRODAS simulation
- 0 muzzle

INTRODUCTION

"A conventional gun is essentially a heat engine in which the propellant contained or injected in the gun chamber is ignited and combusted, transferring its chemical energy into kinetic energy of the projectile" (Maag and Klingenberg, 1996).

The propellant grain geometry influences the gas mass flux supplied by the propellant combustion to impel the projectile. According with the burning area evolution along of the time, the propellant grain is classified as progressive, regressive, or neutral, indicating that the burning area, respectively, increase, decrease or does not change during the burning process (Maag and Klingenberg, 1996, AMCP 706-150, 1965, Farrar and Leeming, 1983, Miner, 2012).

"Multiplex charges consist of two or more different grains of propellants. [...] different grains

can be made to tailor the pressure-time curve while keeping the maximum pressure within the required limit" (Maag and Klingenberg, 1996).

Experiments with duplex charges (combining two types of grains) and triplex charges (combining three types of grains) have shown the viability of multiplex charges in a rifle 7.62 (Culver, 1972).

Lumped parameters models of the interior ballistics have been used to design: propellants, ammunitions and guns (Cronemberger et al. 2014, AMCP 706-150, 1965, Farrar and Leeming, 1983, Miner, 2012, Baer and Frankle, 1962, Gonzalez, 1990, Li and Zhang, 2011, Cheng and Zhang, 2012, STANAG 4367, 2012, Yildirim, Tiryaki and Kabak, 2012, Sadek et al. 2014). These models have low computational cost, which is relevant in stochastic optimization problems.

Optimizations of propellant charges have been conducted to improve the interior ballistics performance (Gonzalez, 1990, Li and Zhang, 2011, Cheng and Zhang, 2012, STANAG 4367, 2012, Yildirim, Tiryaki and Kabak, 2012, Sadek et al. 2014, Li and Zhang, 2012). These optimizations have been used stochastic methods, as PSO and Genetic Algorithm, and deterministic methods, as Augmented Lagrange Multipliers. The design variables have usually been the web thickness and the propellant mass. Nevertheless, deterrent thickness and other geometric parameters of the propellant grains were also optimized. A duplex charge combining seven perforated and single perforated grains was analyzed (Gonzalez, 1990, Sadek et al. 2014) and a triplex charge with seven perforated, single perforated and chord grains was also studied (Gonzalez, 1990).

The present work is devoted to multiplex charge optimization based on interior ballistics simulations by using stochastic optimization methods. Multiplex charges are optimized, combining propellant grains with different burning behaviors: ball (regressive burning), single perforated (neutral burning) and single perforated with inhibition in the outer surface (progressive burning). The stochastic optimization methods PSO and Differential Evolution (DE) are employed and a Lumped Parameters Model (LPM) is used to simulate the interior ballistics of 155 mm howitzer.

LUMPED PARAMETERS MODEL (LPM)

The proposed LPM is based on the projectile dynamics and on the thermodynamics of the gas provided by the propellant burning (Cronemberger et al. 2014, Farrar and Leeming, 1983, Baer and Frankle, 1962).

The propellant burning follows the Robert's law

$$\frac{\mathrm{df_j}}{\mathrm{dt}} = -\frac{\mathrm{B}}{\mathrm{D_j}} \mathrm{P^a} \tag{1}$$

where f is the web fraction, B is the burning rate constant, a is the pressure index, D is the web thickness and the subscript j indicates the type of propellant grain.

The propellant form function, z_j , depends on the grain geometry, following

$$z_1 = 1 - f_1^3 \,, \tag{2}$$

$$z_2 = 1 - f_2, (3)$$

$$z_3 = 1 - \frac{(IR + D_3)^2 - [IR + (1 - f_3)D_3]^2}{(IR + D_3)^2 - IR^2}.$$
 (4)

In Eq.(2-4), z_1 is the form function of the ball grain (regressive burning grain), z_2 is the form function of the single perforated grain (neutral burning grain), z_3 is the form function of the single perforated grain with inhibition in the outer surface (progressive burning grain) and IR is the inner radius of the progressive burning grain.

Therefore, the total mass of gas supplied by propellant burning is

$$m_{g} = \sum m_{p,j} z_{j} \tag{5}$$

where $m_{p,j}$ is the initial mass of the propellant grain of type j.

The Nobel-Abel equation of state is

$$P\left(Vol_{g} - m_{g} c\right) = m_{g} R T$$
 (6)

c, Vol_g , R, m_g and T are, respectively, the co-volume, the volume, the constant, the mass and the absolute temperature of the gas (Maag and Klingenberg, 1996, Cronemberger et al. 2014, AMCP 706-150, 1965, Farrar and Leeming, 1983, Miner, 2012, Baer and Frankle, 1962, Gonzalez, 1990).

The 1st law of the thermodynamics follows

$$Q = \Delta U + W + E_{lost}, \qquad (7)$$

$$Q = \frac{m_g F}{\gamma - 1} + E_i, \qquad (8)$$

$$W = 0.5 \,\mathrm{m}_{\mathrm{proj}} \,\mathrm{V}^2,$$
 (9)

$$\Delta U = \frac{P}{\gamma - 1} \left(Vol_g - m_g c \right). \tag{10}$$

The heat released by igniter and propellant burning, Q, is changed into the projectile kinetic energy, W, into the variation of the gas internal energy, ΔU , and into lost energy, E_{lost} .

In these equations, F is the propellant impetus, γ is the gas specific heats ratio, E_i is the igniter energy and m_{proj} is the projectile mass.

To evaluate the gas volume, it is necessary to consider the initial gas volume in the combustion chamber, the projectile movement and the propellant burning. The volume of gas is

$$Vol_g = Vol_{cc} - \frac{m_p}{\rho} + \pi \frac{d^2}{4} S + \frac{m_g}{\rho}$$
, (11)

where ρ is the propellant density.

The pressure function follows

$$P = \frac{(\gamma - 1)(Q - W - E_{lost})}{Vol_g - m_g c}.$$
 (12)

The translational dynamics of the projectile is described by the Newton's 2nd law,

$$m_{\text{proj}} \frac{dV}{dt} = \pi \frac{d^2}{4} (P - P_{\text{atm}} - P_r). \tag{13}$$

The atmospheric pressure, P_{atm} , and the resistance pressure, P_{r} , act against the projectile movement.

The projectile velocity definition is

$$\frac{\mathrm{dS}}{\mathrm{dt}} = \mathrm{V} \ . \tag{14}$$

Special attention should be devoted to P_r and E_{lost} modelling, since they cannot be directly measured. In the present model, P_r and E_{lost} are spline functions of the projectile velocity furnished by the interior ballistics simulation of the 155 mm howitzer performed with the commercial software PRODAS by using 155M549 input data of the PRODAS library. This simulation furnishes discrete values of P_r , but not E_{lost} data. Then, the PRODAS output data are introduced in Eq. (7) to compute discrete values of E_{lost} . In sequence, splines (by using interp1 SciLab function) are applied on these discrete values of E_{lost} and V, defining the E_{lost} function. The same procedure is applied on P_r and V, defining the P_r function.

A Runge-Kutta fourth order method is implemented in SciLab to solve the system of Eqs.(1), (13) and (14). The initial conditions of the problem are $f_j = 1$, V = 0 and S = 0. The initial pressure is equal to the atmospheric pressure and the initial volume fraction of the burned propellant is null.

155M549 SIMULATIONS

Simulations of a howitzer 155 with M549 ammunition (155M549) were performed by using

PRODAS and LPM, whose input data are shown in Tab. 1.

The form function in the 155M549 LPM simulation follows

$$z = (1-f)(1+0.12f),$$
 (15)

representing a seven-perforated grain (Farrar and Leeming, 1983).

Table 2 summarizes the output data of the simulations performed by PRODAS and LPM. In this table, P_{max} is the maximum pressure, P_0 is the muzzle pressure, t_0 is the total time of the interior ballistics, V_0 is the projectile muzzle velocity and η is the thermal efficiency of the weapon, which can be computed by

$$\eta = \frac{0.5 \text{ m}_{\text{proj}} \text{ V}_0^2}{\frac{\text{F} \sum \text{m}_{\text{p,j}}}{\gamma - 1} + \text{E}_i}.$$
 (16)

Table 1. 155M549 input data.

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PARAMETER	VALUE	
Igniter energy (E _i)	375745.8 J	
Mass of propellant (m _p)	11.7934 kg	
Impetus of the propellant (F)	1.09 MJ/kg	
Ratio of specific heats of the	1.25	
gas (γ)	1.23	
Co-volume (c)	1.0575 10 ⁻³ m ³ /kg	
Propellant grain	seven perforated	
Density of the propellant (ρ)	1660.8 kg/m^3	
Pressure index of the	0.67	
propellant (a)	0.07	
Burning rate constant of the	5.42 10 ⁻⁷ (m/s) Pa ^{-a}	
propellant (B)	3.42 10 (III/S) 1 a	
Web thickness (D)	2.0015 mm	
Mass of projectile (m _{proj})	43.5448 kg	
Projectile diameter (d _{proj})	meter (d _{proj}) 155 mm	
Barrel length (L)	5.08 m	
Volume of the	0.01886 m^3	
combustion chamber (Vol _{cc})		

Tab. 2 reveals the congruence of LPM and PRODAS simulations.

Table 2. 155M549 output data.

Method	η	P _{max} (MPa)	P ₀ (MPa)	t ₀ (ms)	V ₀ (m/s)
PRODAS	0.29	278.8	82	24	826
LPM	0.28	279.7	82	24	823

OPTIMIZATION PROBLEMS

The interior ballistics optimization problems follow

Minimize
$$g(D_j, m_{p,j}),$$
 (17)

subject to
$$0.5 \text{ mm} \le D_j \le 3 \text{ mm}$$
, (18)

$$0 \text{ kg} \le m_{p,1} \le 7 \text{ kg}$$
, (19)

$$0 \text{ kg} \le m_{p,2} \le 11.7934 \text{ kg}$$
, (20)

$$0 \text{ kg} \le m_{p,3} \le 16 \text{ kg}$$
, (21)

$$\sum m_{p,j} \le 31 \,\mathrm{kg} \,, \tag{22}$$

$$P \le P_{\text{max,ref}}$$
, (23)

$$V_0 \ge V_{0,ref} . \tag{24}$$

where the subscript ref indicates the data of the PRODAS simulation in Tab. 2.

The box constraints limits in Eq.(19-21) were chosen to avoid $P>P_{max,ref}$, considering the cases with a single type of propellant grain, and the inequality limit value in Eq.(22), considering the maximum capacity of the combustion chamber.

The 1st optimization problem (1st OP) maximizes the muzzle velocity,

$$g(D_{j}, m_{p,j}) = V_{0}^{-1} + +10^{2} |P_{max} - P_{max,ref}| step(P_{max,ref})$$
 (25)

and the 2nd optimization problem (2nd OP) minimizes the maximum pressure,

$$g(D_{j}, m_{p,j}) = \frac{P_{max}}{P_{max,ref}} + \frac{10^{2} |V_{0} - V_{0, ref}| \left[step(0) - step(V_{0, ref}) \right]}{(26)}$$

In Eq.(25) and (26) step is the unit step function and the second term of these equations are penalty functions. In Eq.(25), the penalty function prevent maximum pressures greater than the 155M549 one (Tab. 2) and, in Eq.(26), the penalty function keep the muzzle velocity greater than the 155M549 one (Tab. 2).

STHOCASTIC OPTIMIZATION METHODS

To solve the optimization problems two methods are employed: Particle Swarm Optimization (PSO) and Differential Evolution (DE). In both methods, the initial population of possible solutions is created inside the feasible domain and each possible solution of the population is called particle.

PSO is based on a flock of birds looking for food (Cheng and Zhang, 2012, Sadek et al. 2014, Colaço et al. 2006) and the evolution rules of the particles can be written as

$$\begin{aligned} y_{n}^{k+1} &= \alpha \ y_{n}^{k} + \beta_{1} \ \text{rand}_{1,n} \Big(x_{b,n}^{k} - x_{n}^{k} \Big) + \\ &+ \beta_{2} \ \text{rand}_{2,n} \Big(x_{bg}^{k} - x_{n}^{k} \Big) \end{aligned} \tag{27}$$

$$x_n^{k+1} = x_n^k + y_n^k. (28)$$

where x is a particle, y is the particle velocity, $\operatorname{rand}_{1,n}$ and $\operatorname{rand}_{2,n}$ are random numbers with normal distribution in the range from 0 to 1, the subscript b indicates the best value of the particle n along its history and the subscript bg indicates the best particle of the population. Moreover, k is the iteration. In PSO, α , β_1 and β_2 are arbitrary values and, in this study, they are, respectively, equal to 0.5, 0.7 and 1.25.

DE is based on the Darwin's Theory of Evolution (Colaço et al. 2006). During the population evolution mutations and chromosome crossover occur. The evolution rule is

$$x_n^{k+1} = \delta_1 x_n^k + \delta_2 \left[x_{bg}^k + CF \left(x_{r1}^k - x_{r2}^k \right) \right], \tag{29}$$

$$\delta_{1} = \begin{cases} 0, & \text{if } \operatorname{rand}_{1,n}^{k} < CR \\ 1, & \text{if } \operatorname{rand}_{1,n}^{k} \ge CR \end{cases}, \tag{30}$$

$$\delta_2 = \begin{cases} 1, & \text{if } \operatorname{rand}_{1,n}^k < CR \\ 0, & \text{if } \operatorname{rand}_{1,n}^k \ge CR \end{cases}$$
 (31)

It takes into account the best particle of the population and two particles x_{r1} and x_{r2} , which are randomly chosen within the population. CR and CF are arbitrary parameters in the range from 0.5 to 1.0, which control crossover and mutation processes. In this work, CR = 0.5 and CF = 0.5.

During the population evolution, particles can fall outside the feasible domain. Thus, repair procedures must put these particles inside the feasible domain (Helwig, 2010). If a box constraint is not satisfied, the particle is put on nearest constraint limit and if Eq.(22) is not satisfied, the particle is discharged and a new particle in the feasible domain is randomly created (Helwig, 2010).

The stopping criterion establishes that

$$\left| \frac{g(D_{j}^{k}, m_{p,j}^{k}) - g(D_{j}^{k-1}, m_{p,j}^{k-1})}{g(D_{j}^{k-1}, m_{p,j}^{k-1})} \right| \le 0.01,$$
(32)

must be satisfied during 50 consecutive iterations.

RESULTS

The PRODAS simulation, reported in Tab. 2, is the reference curve in Fig. 1 to 6. In these figures, optimized propellant charges results computed with PSO and DE, considering populations with 20 and 40 particles, are shown.

Solutions for 1st optimization problem are presented in Fig. 1 to 3.

Fig. 1 reports the pressure inside the weapon along the projectile travel position. It is important to note that increasing the area below the P x S curve, the work and the muzzle velocity increase.

The effect of the population size is more relevant on DE results than on PSO ones. Moreover, the PSO have increased more the area below P x S curve than DE.

Fig. 2 shows the projectile velocity along the projectile travel position. In this figure, it is observed that the highest muzzle velocity was obtained by using PSO with 40 particles in the population. However, every optimized charge has improved the muzzle velocity in relation to the reference case.

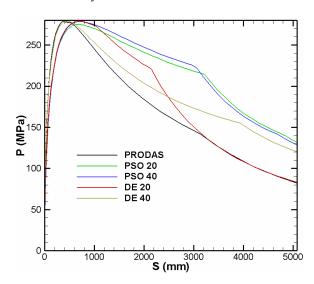


Figure 1. Pressure inside the howitzer (1st OP).

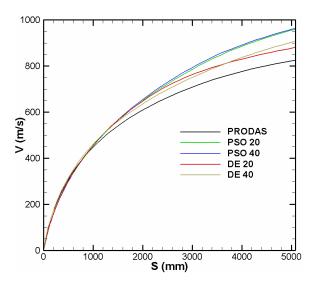


Figure 2. Projectile velocity (1st OP).

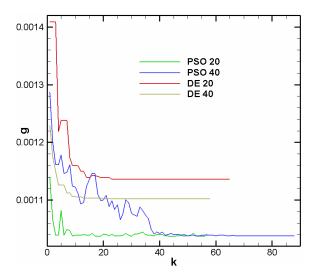


Figure 3. Objective function evolution (1st OP).

Fig. 3 depicts the objective function evolution along of the iterations. PSO have minimized more the objective function than DE. Otherwise, PSO with 40 particles in the population have spent more iterations to converge.

Table 3 summarizes the 1st optimization problem results. The results reveal that optimized charges have large amount of progressive burning grains. Nevertheless, small quantities of regressive and neutral burning grains can significantly change the muzzle velocity. Moreover, PSO 40 is the best choice to improve the muzzle velocity. The PSO 40 muzzle velocity is 17% greater than the PRODAS muzzle velocity. Besides, PSO 40 has the same thermal efficiency value of the reference case (Tab. 2).

Table 3. Summary of 1st OP results.

Method	PSO		DE	
Population	20	40	20	40
η	0.28	0.29	0.31	0.26
k	57	88	65	58
V_0 (m/s)	960	963	880	907
P _{max} (MPa)	275.1	278.8	278.8	278.8
$m_{p,1}$ (kg)	0.000	0.008	0.309	1.502
$m_{p,2}$ (kg)	0.358	0.051	1.697	2.758
$m_{p,3}$ (kg)	16.000	16.000	10.197	11.428
$D_1(m)$	0.0013	0.0028	0.0025	0.0028
$D_2(m)$	0.0029	0.0021	0.0013	0.0026
D ₃ (m)	0.0020	0.0020	0.0017	0.0021

The DE 20 case (Tab. 3) provided the worst optimized muzzle velocity, but the best thermal efficiency. Therefore, it is possible to improve the thermal efficiency with optimized multiplex charges.

Results of the 2nd optimization problem are presented in Fig. 4, 5 and 6 and in Tab 4. These figures are, respectively, analogous to Fig. 1, 2 and 3.

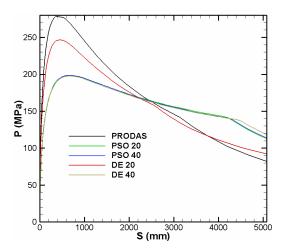


Figure 4. Pressure inside the howitzer (2nd OP).

Fig. 4 and 5 shows that PSO 20, PSO 40 and DE 40 have almost the same maximum pressure and projectile velocity curve.

The evolution of the objective function in Fig. 6 shows that DE 20 is the worst optimized solution. Nevertheless, DE 40 is the best solution, reaching the lowest value for the objective function and spending less iterations to converge.

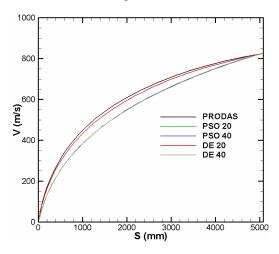


Figure 5. Projectile velocity (2nd OP).

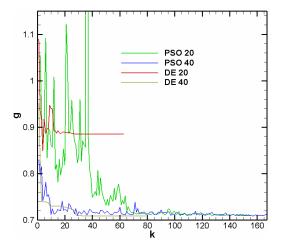


Figure 6. Objective function evolution (2nd OP).

Table 4. Summary of 2nd OP results.

Method	PSO		DE	
Population	20	40	20	40
η	0.24	0.23	0.26	0.23
k	155	167	63	79
V ₀ (m/s)	826	826	826	826
P _{max} (MPa)	197.8	198.6	246.7	197.6
$m_{p,1}$ (kg)	0.004	0.026	0.000	0.000
m _{p,2} (kg)	0.006	0.066	6.455	0.000
m _{p,3} (kg)	14.383	14.369	6.725	14.829
D ₁ (m)	0.0007	0.0019	0.0010	0.0005
D ₂ (m)	0.0030	0.0015	0.0017	0.0011
D ₃ (m)	0.0022	0.0022	0.0023	0.0023

Table 4 summarizes the results for the 2nd optimization problem. The thermal efficiencies reported in Tab. 4 are worse than reference case thermal efficiency (Tab. 2), but the same muzzle velocity is attained. Moreover, DE 40 maximum pressure is 29% lower than reference case maximum pressure. Noteworthy that DE 40 propellant charge has only progressive grains.

CONCLUSION

The present LPM has proposed an alternative approach to evaluate the resistance pressure and the lost energy, employing output data of PRODAS simulation. The congruence between LPM and PRODAS simulations has shown that such alternative approach is a promising procedure.

The results of optimized propellant charges have shown that large amounts of progressive burning grains should be used in multiplex charges. In this sense, single perforated grains with inhibition in the outer surface were useful to increase the muzzle velocity and to minimize the maximum pressure. Moreover, small amounts of regressive and neutral propellants grains are relevant to increase the powder load density, being useful to increase the muzzle velocity. On the other hand, regressive and neutral grains should be avoided to reduce the maximum pressure inside the gun.

Increasing the muzzle velocity, the utility range of the weapon is also increased and reducing the maximum pressure of the interior ballistics, lighter weapons could be designed, improving the mobility.

Particle Swarm and Differential Evolution were effective to solve the optimization problems. Though, PSO results were better than DE ones, maximizing the muzzle velocity, and DE results were better than PSO ones, minimizing the maximum pressure.

The size of populations in the studied stochastic methods has influenced the optimized results. The quality of the results was improved, when the population size was increased. However, the effects of the population size on PSO results were less relevant than on DE ones.

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