Development of a machine learning model to determine the forces on the piston in the pump-tube of a two-stage gas gun deforming due to a taper

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Abstract

Abstract here.

1 TODO

- 1. Get output parameters
- 2. Model automation
- 3. Mesh optimization

2 Introduction

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• We will use: ton, mm, s, N, MPa, N-mm

2.1 Gas gun design

2.1.1 title

3 Finite element model

A finite element model of the scenario is created to run computation experiments. These experiments provide data for the force and plastic deformation that the cap experiences as it makes contact with the transition piece. After several computational experiments are run, the resulting data is used to train a machine learning model that can predict the forces experienced by the cap without the need for a finite element model.

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MASS	LENGTH	TIME	FORCE	STRESS	ENERGY	Steel Density	Steel Modulus	G - Gravity Constant
kg	m	s	N	Pa	Joule	7.83E+03	2.07E+11	9.81
kg	mm	ms	kN	Gpa	kN-mm	7.83E-06	2.07E+02	9.81E-03
g	cm	s	dyne	dyne/cm^2	erg	7.83E+00	2.07E+12	9.81E+02
					1e7 N-			
g	cm	us	1e7N	Mbar	cm	7.83E+00	2.07E+00	9.81E-10
g	mm	s	1e-6N	Pa	1e-9 J	7.83E-03	2.07E+11	9.81E+03
g	mm	ms	N	Mpa	N-mm	7.83E-03	2.07E+05	9.81E-03
ton	mm	s	N	Mpa	N-mm	7.83E-09	2.07E+05	9.81E+03
lbf-								
s^2/in	in	s	lbf	psi	lbf-in	7.33E-04	3.00E+07	3.86E+02
slug	ft	S	lbf	psi	lbf-ft	1.52E+01	4.32E+09	32.2

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3.1 Geometry

Axial symmetry is used to reduce the problem to two dimensions. The resulting geometric model is presented in Figure 2. Explain the model and the parts and geometries Table of geometric values

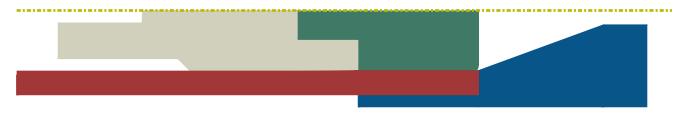


Figure 2: Geometric model of two-stage gas gun pump tube.

3.2 Material models

The material models used for each component are presented here.

3.2.1 High density polyethylene (HDPE)

High Density Polyethylene (HDPE) is modelled using the Ramberg-Osgood model as done so in the literature [2]. In 1D, the model is given by

$$E_c \varepsilon = \sigma + \alpha_c \sigma \left[\frac{|\sigma|}{\sigma_{yc}} \right]^{n-1}, \tag{1}$$

where E_c is the Young's modulus, α_c is the yield offset, σ_{yc} is the yield stress and n is the hardening exponent. In 3D, the model has the additional parameter of Poisson's ratio, denoted by ν_c . The material parameters are determined by fitting the model to data from [2] using a Python script which can be found at this link. Additionally, a 3D uniaxial Abaqus simulation utilizing the model is compared to the 1D model (equation (1)) to verify that the model behaves as expected in Abaqus. The results of the simulation, the behaviour of the model, and the data are displayed in Figure 3. The calibrated material parameters, as well as the other parameters required to define the model, are presented in Table 1.

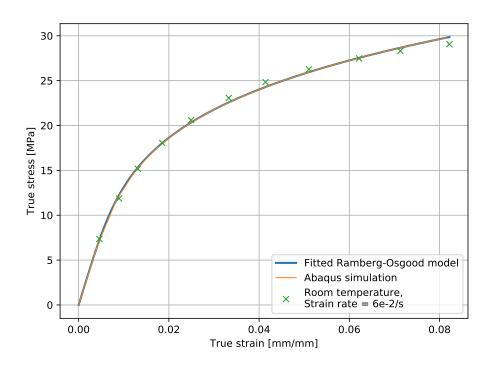


Figure 3: Stress-strain relation for HDPE modelled with the Ramberg-Osgood model. The model is calibrated to fit data from [2] (denoted by green 'x's). Additionally, the 1D model is compared to a uniaxial ABAQUS simulation using the calibrated material parameters. It is clear that the curves match each other and provide a good fit to the data.

Table 1: Material parameters for HDPE

Density ρ_c	Young's	Poisson's ra-	Yield stress	Yield offset	Hardening
ton/mm_3 [3,	modulus E_c	tio ν_c [7]	σ_{yc} MPa	α_c	exponent n
2]	MPa				
0.95×10^{-9}	1.63×10^{3}	0.46	14.6	0.342	4.26

3.2.2 Mild steel

The mild steel components are expected to remain in the elastic region and so only linear elastic material behaviour is considered. The material parameters are given in Table 2.

Table 2: Material parameters for mild steel Ref later

Density ρ_t	Young's	Poisson's ra-
$\mathrm{ton/mm_3}$	modulus E_t	tio ν_t
	MPa	
8×10^{-9}	200×10^{3}	0.3

3.2.3 Aluminium

The aluminium components are expected to remain in the elastic region and so only linear elastic material behaviour is considered. The material parameters are given in Table 3.

Table 3: Material parameters for aluminium taken from [1]

Density ρ_p	Young's	Poisson's ra-
$\mathrm{ton/mm_3}$	modulus E_p	tio ν_p
	MPa	
2.69×10^{-9}	68.3×10^{3}	0.34

3.3 Interactions

Contact interactions and tie constraints are applied between various parts in the assembly, see Figure 2. The relevant coefficients of friction for these interactions are obtained from the literature and summarised in Table 4.

Table 4: Relevant coefficients of friction for problem

HDPE on steel μ_{cs} [6]	steel on steel μ_{ss} [8]	aluminium on steel μ_{as} [5]
0.25	0.2	0.5

The cap is assigned a contact interaction with the pump tube and the transition piece. Hard contact is used to describe the normal behaviour and sliding friction is used to describe the tangential behaviour with a coefficient of friction of μ_{cs} , see Table 4. The piston is similarly assigned a contact interaction with the pump tube with a coefficient of friction of μ_{as} .

The pump tube is tied to the transition piece to approximate the threaded interaction. Additionally, a tie constraint is assigned between the cap and the piston. The interaction between the cap and piston would likely better be described by cohesive interaction. However, the interaction between the cap and the piston is expected to have little influence on the results of interest from the model, and so a tie constraint is used as a convenient first approximation.

3.4 Initial and boundary conditions

- 1. Region A and B are encastracted
- 2. The piston and end cap are assigned an initial velocity of v_0
- 3. A pressure load is applied to the back of the piston with varying magnitude throughout time.
- 4. A pressure load is applied to the front of the cap with varying magnitude throughout time.

3.5 Time integration

An implicit Newmark-method is used with time increments of Δt

3.6 Variable input parameters and desired output parameters

Input parameters to be varied:

- 1. v_0
- 2. p_{back}
- 3. p_{front}

Outputs of interest: 1. v $2. x_t$ 3. $\langle \Phi_c \rangle$ 4. Δp Outputs to later be predicted: 1. F_c 2. $\left\langle \dot{\Phi}_{c}\right\rangle$ Single test simulation 3.7 Mesh optimization 3.8 Parameters 1. ratio of element expansion for piston 2. ratio of element expansion for tube 3. n elements Objective function 1. Min elements Constraint 1. Force 2. Dissipation Machine learning surrogate model 4 Feature engineering 4.1 Predictive features: 1. Coefficient of friction: μ 2. Taper angle: α 3. Velocity: v4. Distance between piston front and taper start: x_{taper} 5. Pressure difference between piston front and back: Δp

6. Piston length: l_p

7. Piston density: ρ_p

8. Accumulative plastic strain in the piston: γ

Dependent variables:

- 1. Axial force on piston due to taper: F_z
- 2. Increment in accumulated plastic dissipation: $\Delta \gamma$

4.1.1 Dimensional analysis

Table 5: Fundamental dimensional units

Table 6: Variables on which the problem depends Check that units are correct

Component	Description	Symbol	Units
Cap	Axial contact force	F_c	$[MLT^{-2}]$
	Rate of volume average plastic dissipation	$\left\langle \dot{\Phi}_{c} ight angle$	$[ML^{-1}T^{-3}]$
	Volume average plastic dissipation	$\langle \Phi_c angle$	$[ML^{-1}T^{-2}]$
	Pressure difference	ΔP	$[ML^{-1}T^{-2}]$
	Velocity	v	$[LT^{-1}]$
	Distance of cap past taper	x_t	[L]
	Length past piston	l_c	[L]
	Diameter	d_c	[L]
	Density	$ ho_c$	$[ML^{-3}]$
	Young's modulus	E_c	$[ML^{-1}T^{-2}]$
	Poisson's ratio	$ u_c$	[ullet]
	Yield stress	σ_{yc}	$[ML^{-1}T^{-2}]$
Piston	Length	l_p	[L]
	Diameter	d_p	[L]
	Density	$ ho_p$	$[ML^{-3}]$
	Young's modulus	E_p	$[ML^{-1}T^{-2}]$
	Poisson's ratio	$ u_p$	[•]
Transition piece	Length	l_t	[L]
	Diameter change	Δd_t	[L]
	Density	$ ho_t$	$[ML^{-3}]$
	Young's modulus	E_t	$[ML^{-1}T^{-2}]$
	Poisson's ratio	$ u_t$	[ullet]
Friction	Cap-on-steel CoF	μ_{cs}	[•]
	Steel-on-steel CoF	μ_{ss}	[ullet]
	Aluminium-on-steel CoF	μ_{as}	[ullet]

$$\Pi = F_c^{k_1} \left\langle \dot{\Phi}_c \right\rangle^{k_2} \Delta P^{k_4} v^{k_5} \left\langle \Phi_c \right\rangle^{k_3} x_t^{k_6} \tag{2}$$

$$r = 3 \tag{3}$$

Table 7: Non-dimensional parameters

$$F_c \quad \left\langle \dot{\Phi}_c \right\rangle \quad \left\langle \Phi_c \right\rangle \quad \Delta P \quad v \quad x_t$$

$$k_1 \quad k_2 \quad k_3 \quad k_4 \quad k_5 \quad k_6$$

$$n = 24 \tag{4}$$

$$n_{\text{n-dim}} = n - r = 21 \tag{5}$$

- 1. Enter units as symbols in sympy
- 2. Construct dimensional matrix
- 3. Find solutions k_i^j to dimensional matrix by n-r values in k to 0 (except for 1 which on sets to 1).
- 4. Automatically print out resulting dimensionless products table.

Based on [4]:

Notes:

1.

Direct quote from [4]

"Dimensional analysis is a method with the aid of which one may for instance test a formula for dimensional correctness. It leads to a first understanding of the solution of a physical problem and yields a precise information about the number of variables that are necessary to describe it, a fact that is particularly important when experiments are being performed. Very often dimensional analysis reduces the number of variables upon which a physical problem was initially surmised to depend. If for instance the quantity y depends upon $x_1, x_2, ..., x_n$, where all quantities have a certain physical dimension, then dimensional analysis shows that y can only depend upon certain products of powers of $x_1, x_2, ..., x_n$, a fact that corresponds regularly to a considerable reduction of the number of variables. Naturally then, experiments may more simply or more economically be performed than without knowledge of this fact."

"The first step in a dimensional analysis consists in the listing of the parameters, which influence a physical problem. This step is very decisive. If too many variables are listed that may describe a physical problem, then the final equations will contain superfluous variables, if too few variables are introduced, incomplete equations may emerge, which results in incomplete equations or "more often" false inferences or the result can not be expressed in terms of dimensionally homogeneous functions."

4.2 Experimental input parameters

In order of importance:

- 1. Initial velocity: v_0
- 2. Pressure path: p_{path}
- 3. Piston length: l_p
- 4. Coefficient of friction: μ

4.3 Experimental results

- 4.4 Model
- 5 Packaging of model for use in 1D code
- 5.1 PIP
- 5.2 Usage example

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