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myBWID version 1.0

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1. Introduction

1.1 Scope

This document describes in brief the usage of myBWID version 1.0. The Bouc-Wen model is described in chapter 2. Next, the main features of the program are presented through a number of selected examples.

1.2 Program requirements

The minimum requirements are:

- Operating System: Microsoft® Windows /NT/2000/XP/Vista
- > Visual Basic 6 Service Pack 5 runtime libraries.

1.3 Abbreviations

SDOF:	Single Degree of Freedom
ODE:	Ordinary Differential Equation

1.4 About myBWID (my Bouc-Wen IDentification) version 1.0

This program employs evolutionary algorithms for the identification of SDOF Bouc-Wen hysteretic systems. Apart from hysteretic damping, the identification may include viscous type effects. The program was developed and used for research purposes. The main results of this research were published recently [1].

The hybrid evolutionary algorithm described in [1] is implemented within the program. However, other Genetic Algorithms (GAs) including standard GA, microGA, Sawtooth GA [5] etc can also be used. In addition, a greedy ascent hill climber (GAHC) may be employed which greatly improves performance.

Some important notes:

• The software is provided "as is". Make sure to read the terms of usage in the about form.

- At this moment, this manual provides only a small number of guidelines on how to use the software. Future releases of this manual will include the description of more program features.
- The performance of all algorithms described in [1] was measured with respect to the number of function evaluations, not the time needed to conclude the analysis. The program has *not* been designed to provide the fastest execution possible. A number of system parameters have been set in such a way so as to facilitate debugging, although they may entail a significant overhead. These include the frequent redrawing of the response of the current best result and the output of a complete analysis report which describes all stages of execution.
- If you are interested in this work or require further information, do not hesitate to contact professor V.K. Koumousis (<u>vkoum@central.ntua.gr</u>) or A.E. Charalampakis (<u>achar@mail.ntua.gr</u>)

2. Bouc Wen Model

2.1 General

The Bouc – Wen model is a smooth hysteretic model which is very popular because of its versatility and simplicity. It was first introduced by Bouc in 1967 [2]. In 1976, Wen [3] extended the model and demonstrated its versatility by producing a variety of hysteretic patterns.

2.2 Formulation

According to Bouc-Wen model, the restoring force of a SDOF system can be written as:

$$F(t) = a \cdot \frac{F_y}{u_y} \cdot u(t) + (1-a) \cdot F_y \cdot z(t)$$
(7.2.1)

where, F_y is the yield force, u_y is the yield displacement, a is the ratio of postyield to pre-yield (elastic) stiffness and z(t) is a dimensionless hysteretic parameter obeying a single differential equation with zero initial condition:

$$\dot{z}(t) = \frac{1}{u_{y}} \left[A - \left| z(t) \right|^{n} \cdot \left(\gamma \cdot sign(\dot{u}(t) \cdot z(t)) + \beta \right) \right] \cdot \dot{u}(t)$$
(7.2.2)

where, A, β, γ, n are dimensionless quantities controlling the shape of the hysteresis loop.

The equation of motion for a SDOF system with linear viscous damping c is given as:

$$m \cdot \ddot{u}(t) + c \cdot \dot{u}(t) + F(t) = f(t)$$
(7.2.3)

where, u(t) is the displacement, F(t) is the restoring force, f(t) is the excitation force. Substituting (7.2.1) into (7.2.3) we obtain:

$$m \cdot \ddot{u}(t) + c \cdot \dot{u}(t) + a \cdot \frac{F_y}{u_y} \cdot u(t) + (1 - a) \cdot F_y \cdot z(t) = f(t)$$
(7.2.4)

Equations (7.2.2) and (7.2.4) are transformed into a state-space form as follows:

$$\begin{cases} x_{1}(t) = u(t) \\ x_{2}(t) = \dot{u}(t) \\ x_{3}(t) = z(t) \end{cases}$$
(7.2.5)

$$\begin{cases} \dot{x}_{1}(t) \\ \dot{x}_{2}(t) \\ \dot{x}_{3}(t) \end{cases} = \begin{cases} x_{2}(t) \\ -\frac{1}{m} \cdot \left[c \cdot x_{2}(t) + a \cdot \frac{F_{y}}{u_{y}} \cdot x_{1}(t) + (1-a) \cdot F_{y} \cdot x_{3}(t) - f(t) \right] \\ \frac{1}{u_{y}} \cdot \left[\left(A - \left| x_{3}(t) \right|^{n} \cdot \left(\gamma \cdot sign(x_{2}(t) \cdot x_{3}(t)) + \beta \right) \right) \cdot x_{2}(t) \right] \end{cases}$$
(7.2.6)

The above system of three first order non-linear ODEs is solved numerically following Runge-Kutta $4^{th} - 5^{th}$ order or Livermore stiff ODE integrator which is based on a "predictor-corrector" scheme [4].

2.3 Parameter Constraints

A number of parameter constraints are necessary. In particular, A = 1 and $\beta + \gamma = 1$ should be imposed for reasons of physical consistency of the model [1]. By default, these constraints are active in the program. You can choose to deactivate the constraints by using the check boxes in the "GA" frame of the main window. However, this is not recommended.

3. Example 1: El Centro excitation

3.1 General

The usage of this program will be described through examples. The first example refers to a SDOF system under the El Centro excitation.

3.2 Excitation

The first step is to define the excitation. Click the "Load from file" button at the top left corner:

myBWID 1.0			
Experiments Load from file Experiment : Weight :	Clear Mass : Time step : Counter : Initial U : Initial Udo ; Initial Z :	Force Controlled Sin Excitation Adv. Sin Excitation Sin Displacement Displacement Controlled Sin Displacement	Display Experiment : Graph X Axis : Graph Y Axis : Craph Y Axis :
Solution Method : C Livemore @ RK4 A : 1.0 beta : 0.1 gamma : 0.9 a : 0.10 c : 0.0000 Fy : 2.86 Uy : 0.111 Noise : 0.00	GA Method:	- gamma = 1 - Boundi - Best result 	About my®WID ling Mode Bounding

Select the "El Centro with mass = 28.6.xpr'' file from the main directory of the program:

Load file	e 🔹 🛛 🔀
Look in: 险) myBWID Setup Directory 💽 🔶 💼 🖝 📰 🗸
Popov 5.	э with mass = 28.6.xpr xpr
File name:	El Centro with mass = 28.6.xpr Open
Files of type:	Experiment files (*.xpr) Cancel

As soon as you load the file, the program will evaluate the response of the model. Consistent unit system is used. The "true" model parameters are provided by the frame entitled "Solution" on the left. When identification is under way, the model

parameters corresponding to the current best result are summarized within the "GA" frame, under the "Best result" label.



3.3 Graph

The program automatically draws the response of both the "true" and "identified" model in a picture box on the right of the main window. The graph of "true" response of the model is drawn using a thick red line. The response of the current best result of identification is drawn using a thin blue line.

You can select the quantity that corresponds to the X and Y axes by using the drop-down lists in the "Display" frame:



3.4 Objective Function

The normalized Mean Squared Error (MSE) of the predicted time history $\hat{y}(t|\mathbf{p})$ as compared to the reference time history y(t) is used as objective function. When cast in discrete form, it can be expressed as:

$$OF(\mathbf{p}) = \frac{\sum_{i=1}^{N} (y(t_i) - \hat{y}(t_i | \mathbf{p}))^2}{N \cdot \sigma_y^2}$$
(1)

where, **p** is the parameter vector, σ_y^2 the variance of the reference time history and *N* the number of points used. The time history of the displacement and external force is used for force- and displacement-controlled experiments, respectively.

3.5 GA Analysis

In order to perform a GA analysis, click the "Options" button:

Solution Mass: 28.6 Force Controlled Force Controle	🕫 myBWID 1.0	
Solve Now Run GA Analysis Options Show Results Displacement	Experiments Experiments Experiments Experiment: 1 of 1 Force-Controlled Response is calculate Weight: 1 Solution Method: C Livermore RK4 A: 10 beta: 01 gamma: 0.3 a: 20 a: 010 c: 0.0000 Fy: 286 Uy: 0.111 Noise: 0.00 Solve Now	Force Controlled Diplay Image tep: 102 Sin Excitation Image tep: 102 Sin Excitation </td

Make the appropriate selections in the options form:

GA Options	×
GA Options Ganeric Data Output report at a specific interval Output reyort at a specific interval Selection Selection Selection Selection Cossover method: Foudette wheel Cossover method: Single Point Cossover method: Single Point Uncessover method: Single Point Uncessover method: Single Point Uncessover method: Output Ou	Finess scaling Scale fitness Linew scaling Cmult coefficient : 2 Population Constant Population Size : 100 Constant Population Size : 100 Saw - Tooth N meen (individuals) : 25 T (period, generations) : 5 D (amplitude, individuals) : 20 Cifleria to stop analysis Maximum runs : 1 Best remains for max generations : 0 Contents to stop analysis Contents for max generations : 0 Compare obj. function value with a target value D and target value Compare obj. function value with a ta
1 millio (24	Ok Cancel

The default GA options include the usage of Sawtooth GA [5] with mean population size equal to 25, amplitude of variation equal to 20 and period 5 generations. In addition, a hill climber is employed when the GA analysis has concluded.

Click the "Set accuracy by divisions" button in the main form:



Select the number of divisions that will be applied to the current parameter ranges, as defined within the "GA" frame. This governs the length of the chromosome. Using 128 ($=2^7$) divisions per parameter results in a chromosome length equal to 7 times the number of parameters to be identified.

By default, the lower and upper boundary of the viscous damping coefficient is equal. This means that the viscous damping coefficient will not be identified.



Hit the "Run GA Analysis" button:

Select "0" as the seed for the random number generator. A progress form appears:

GA Progress		
- GA Info		Display
Current Run:	1	Experiment :
Current Generation:	71	El Centro with
Population size:	45	
r opoiddorr acc.	1.0	Graph X Axis :
Best obj. function value:	2.426451172921	
Individuals analyzed:	1795	
	,	Graph Y Axis :
Abort	1	F 💌

The response of best identified result is displayed in blue in the main window. Hit "Abort" once to abort GA analysis and invoke the hill climber directly. Hit "Abort" again to abort the hill climbing analysis.

Hit "Show Results" in the "GA" frame to open the output file of the last analysis executed:

🖻 myGA.out - Notepad
File Edit Format View Help
myGA analysis started at: 19/3/2008 6:38:54 μμ
Options:
Seed used: -6200
Maximum runs: 1 Elitism between runs: Yes Flite individual: 1
For each run:
Population size variation: Saw-Tooth Mean value N: 25 Period T: 5 Amolitude D: 20
Maximum generations: 3000
Convergence criterion: No
GA: Selection method: Biased Roulette Wheel Niching: No Crossover method: Single Point Probability: .7 Children per crossover: One Jump mutation: No Elitism between generations: Yes Elitism between generations: Yes Elitism between generations: 1 Objective function value :1 Objective function value check: Yes Target objective function value: 0 Stop when best objective function value is LOWER than or EQUAL to target Best individual endurance check: No Output file is appended whenever the best individual changes Objective function value rounding: No
Total chromosome length (bits): 35
Variables:
Variable 1: Title: A Type: Real Locked value: 1
Variable 2: Title: beta Type: Real Locked value: .1

3.6 Bounding Analysis

A new hybrid identification scheme that is based on bounding of parameters was presented recently [1]. The algorithm is implemented within the program. A brief description of the algorithm is presented below; for more information refer to the published paper or contact the authors.

According to the method presented in [1], identification is performed in stages. Initially, very wide feasible ranges of parameters are specified which are guaranteed to include the optimum values. Next, M_r short independent GA analyses are performed. These consist of Sawtooth GA [5] followed by Greedy Ascent Hill Climbing (GAHC) [6]. These analyses can be performed in parallel within a computer network, thus reducing drastically the computational time. The results are gathered, sorted and the worst M_r ones are truncated (ignored).

Considering the remaining $M_r - M_t$ solutions, weight coefficients are assigned as follows:

$$w_k = \frac{\max_k \left(OF_k \right)}{OF_k} \tag{2}$$

where, OF_k is the objective function value of the k^{th} solution. According to the above formulation, the worst solution is assigned a weight of unity. The weighted mean value of parameter *i* is calculated as follows:

$$m_{i} = \frac{\sum_{k=1}^{M_{i}-M_{i}} (w_{k} \cdot p_{ik})}{\sum_{k=1}^{M_{i}-M_{i}} (w_{k})}$$
(3)

where, p_{ik} is the value of parameter *i* of the k^{th} solution. The descriptive weighted standard deviation is expressed as:

$$s_{i} = \sqrt{\frac{\sum_{k=1}^{M_{r}-M_{t}} \left(w_{k} \cdot \left(p_{ik}-m_{i}\right)^{2}\right)}{\sum_{k=1}^{M_{r}-M_{t}} \left(w_{k}\right)}}$$
(4)

The new trial upper and lower bounds of parameter i are defined symmetrically around the weighted mean value m_i as follows:

$$\overline{u}_{i,j+1} = m_i + q \cdot s_i$$

$$\overline{l}_{i,j+1} = m_i - q \cdot s_i$$
(5)

where, q is a real coefficient which represents the semi-width of the new trial range in terms of standard deviations. Finally, the feasible domain of parameter i in the next optimization step is defined as the intersection of the current range with the trial range:

$$\left[l_{i,j+1}, u_{i,j+1}\right] = \left[l_{ij}, u_{ij}\right] \cap \left[\overline{l}_{i,j+1}, \overline{u}_{i,j+1}\right]$$
(6)

This process is repeated until the ratio of the current feasible range to the initial feasible range reaches a specific value (10^{-4} by default) for all model parameters. At this point, the upper and lower bounds of all parameters almost coincide. This consists a very convenient termination criterion.

To perform Bounding analysis, make sure you have loaded the "El Centro with mass=28.6.xpr" file. Then, hit the "Bounding" button:



The following form appears:

Bounding Scheme						×
Total Iterations	Max Number of iterations:	40	Max obj fun. Runs Stat Runs End = Trunc.Stat Trunc.End Gen.Stat Gen.End SDev.Stat Lg.Stat = Lg.End = Ar1 = Ar2 = □ Luse Wi ○ Use Wi ○ Use Wi	100.0 20 20 5 5 3 3 3 3.0 2.0 10 10 10 0.001 0.0001 progress ax R/Ri ean R/Ri	runs runs Sawtooth periods Sawtooth periods bits per parameter bits per parameter	
			UK		Galder	

The following settings can be changed:

- *Runs.Start/End* is the number M_r of independent short GA analyses at the beginning and at the end of identification. A linear progress rule is implemented within the program, as analyzed in [1].
- *Trunc.Start/End* is the number M_t of results that are truncated (ignored).
- *Gen.Start/End* is the number K_T of Sawtooth GA periods that are applied for each run.
- *SDev.Start/End* is the *q* coefficient which signifies the semi-width of the new trial range in terms of standard deviation.
- *Lg.Start/End* is the gene length in bits per unknown parameter.
- *Ar2* is the termination range ratio. When the range ratio of all model parameters reaches this value, analysis stops.
- *Max number of iterations* is the maximum number of optimization steps that are allowed per analysis. Normally, this should never be triggered.
- *Total Iterations* is the total number of analyses that should be performed. This is very helpful when a statistical sample of the algorithm performance is needed.

The other settings should not be modified.

Press "Ok" to start the bounding analysis. Upon conclusion, all results are automatically put into folders within the root directory of the program.

4. Example 2: Identification using real data

4.1 General

The second example refers to a full-scale bolted-welded steel connection that was tested by Popov and Stephen [7]. A Bouc-Wen model will be identified so that its response closely fits the experimental data.

4.2 Excitation

Press the "Clear" button to clear all experiments. Then, click the "Load from file" button at the top left corner:

🗯 myBWID 1.0			
Experiments Load from file Experiment : Weight :	Clear Mass : Time step : Counter : Counter : Innial U : Innial U dot : Innial U dot : Innial Z : Counter :	Force Controlled Sin Excitation Adv. Sin Excitation Sin Displacement Displacement Controller Sin Displacement	d Graph Y Axis :
Solution Mathod : C Livermore G RK4 A: 1.0 beta: 0.1 gamma: 0.9 n: 2.0 a: 0.10 c: 0.0000 Fy: 2.265 Uy: 0.111 Noise: 0.00 Solve Now	BA ▼ A= Method: ▼ Deta C Livermore ▼ Deta From To A 10 Dot 00 beta: 00 10 100 n: 100 n: 100 a: 00 00 100 e: 00 y: 00 MSE for this experime Ubjective function val Dissipated Ener Run GA Anatusis Options	terms terms	About my6W/D Bounding Mode Bounding

Select the "Popov 5.xpr" file from the main directory of the program:

Load file	? 🛛
Look in: 🖆 myBWID Setup Directory 💽 🔶 🖆	-
El Centro with mass = 28.6.xpr	
File name: Popov 5.xpr	Open
Files of type: Experiment files (".xpr)	Cancel

In this case, the response is given by experimental data which are stored in the file. Therefore, the "true" parameter values in the "Solution" frame are inactive.

4.3 Graph

The program automatically draws the response of both the "true" and "identified" model in a picture box on the right of the main window. The graph of "true" response of the model is drawn using a thick red line. The response of the current best result of identification is drawn using a thin blue line.

Since this excitation is based on experimental data, only external force and displacement data are available. Select the "Fext" option in the "Graph Y axis" dropdown list to view the hysteretic loops:



4.4 Feasible domain

Set the proper feasible domain ranges as follows:

- *F_v*: from 0 to 1000000
- u_{y} : from 0.001 to 0.1

These are the initial parameter ranges that were used in [1]. Regarding the other model parameters, the correct initial feasible domains are selected by default.

At this point, you can perform the identification using either a GA variant or the Bounding scheme, as described in the first example. Due to the smaller amount of data, this second example is identified much faster. The use of Bounding scheme with the default settings is recommended.

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