Asymptotic trends in time-varying oscillatory period for a dual-staged torsional system

Michael D Krak and Rajendra Singh

Abstract
The primary goal of this article is to propose a new analysis tool that estimates the asymptotic trends in the time-varying oscillatory period of a non-linear mechanical system. The scope is limited to the step-response of a torsional oscillator containing a dry friction element and dual-staged spring. Prior work on the stochastic linearization techniques is extended and modified for application in time domain. Subsequently, an instantaneous expected value operator and the concept of instantaneous effective stiffness are proposed. The non-linear system is approximated at some instant during the step-response by a linear time-invariant mechanical system that utilizes the instantaneous effective stiffness concept. The oscillatory period of the non-linear step-response at that instant is then approximated by the natural period of the corresponding linear system. The proposed method is rigorously illustrated via two computational example cases (a near backlash and near pre-load non-linearities), and the necessary digital signal processing parameters for time domain analysis are investigated. Finally, the feasibility and applicability of the proposed method is demonstrated by estimating the softening and hardening trends in the time-varying oscillatory period of the measured response for two laboratory experiments that contain clearance elements and multi-staged torsional springs.

Keywords
Discontinuously non-linear systems, impact dynamics, digital signal processing, effective stiffness, impact dynamics, free vibration

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Introduction
Most torque-transmission devices contain discontinuous non-linearities, such as clearances, multi-staged springs, pre-load elements, and stoppers\textsuperscript{1-11}; prime examples of this family are vehicle powertrains and drivelines, which are composed of gears, spline shafts, synchronizers, and clutch dampers. This inherent set of non-linear features gives rise to undesirable vibro-impact phenomena, often known in the ground vehicle industry in terms of gear rattle\textsuperscript{4,9} and vehicle driveline clunk\textsuperscript{5-8} problems. Numerical and experimental studies concerning free vibration or step-response\textsuperscript{5-8} clearly show regime-dependent (say double-sided or single-sided impact\textsuperscript{4}) and time-varying oscillatory periods, as well as distinct asymptotic trends (hardening, softening, or linear).\textsuperscript{4-11} In an effort to better understand previously observed results and assess such non-linear phenomena in measured or numerical motion signatures,\textsuperscript{4-11} this paper will propose a new analysis tool. The chief goal is to estimate the trends in time-varying oscillatory periods of a non-linear torsional oscillator with a dual-staged spring and associated dry friction element. The method will be computationally verified and experimentally validated.

Analytical formulation of the problem
Consider a simple and yet representative time-invariant, non-linear torsional system described by a single degree of freedom (SDOF) model shown in Figure 1(a). The corresponding governing equation of motion is the following
\[ J\ddot{\theta} + h \tanh(\eta\dot{\theta}) + \Phi(\theta) = T(t) \] (1)

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The dynamic response within each stage only may be addressed by a linear sub-system with the corresponding natural periods

\[ \tau_{nj} = 2\pi(J/k_j)^{0.5}, \quad j = I, II \]  (4)

The stiffness of the first stage is scaled such that \( k_1 = \alpha k_{II} \) where \( \alpha \geq 0 \). For the case of backlash non-linearity with \( \alpha = 0 \), stage I is a clearance element and \( \tau_{nI} \rightarrow \infty \). However, when \( \alpha \rightarrow \infty \), stage I is a pre-load feature and \( \tau_{nI} \rightarrow 0 \); the limited case is \( \alpha = 1 \), where \( k(\theta) \) is simply a linear spring. External torque \( T(t) \) is a step applied at \( t = 0 \) and is such that the initial and final operating points \( ((\theta_0, T_0) \) and \( (\theta_f, T_f) \), respectively) lie on different stages, as shown in Figure 1(b). This ensures a comprehensive non-linear response with three distinct regimes and asymptotic trends of the time-varying oscillatory period, as seen in prior work.\(^4\)\(^-\)\(^8\)

Specific objectives are as follows: (1) propose the instantaneous effective stiffness concept and utilize it to estimate the asymptotic trends of the time-varying oscillatory period; (2) propose the necessary digital signal processing parameters for related time domain analyses; and (3) validate the proposed method by estimating the asymptotic trends in the time-varying oscillatory period of two recent laboratory experiments.\(^8\) The scope of this paper is limited to the time domain analysis of the step-response of time-invariant torsional systems. Additionally, it is assumed that damping can be approximated as Coulomb friction, and therefore, the oscillatory period may be approximated by the natural period of the first mode. This article extends the stochastic linearization techniques\(^13\),\(^14\) proposed by Wallaschek\(^15\) and Rook and Singh.\(^16\)

**Time-varying oscillatory period**

The typical step-responses of the example case (Figure 1) is predicted for two illustrative cases, a near backlash non-linearity (\( \alpha = 0.1 \)) and near pre-load non-linearity (\( \alpha = 10 \)). The values of \( k_{II} \) and \( \Theta_I \) are estimated from a production vehicle clutch damper.\(^8\) Torsional inertia \( J \) is sized so that \( \tau_{nII} = 0.1 \) s, which is within the range of the first torsional surge mode of a vehicle driveline.\(^4\)\(^-\)\(^9\) The amplitudes of \( h \), \( (\theta_0, T_0) \), and \( (\theta_f, T_f) \) are judiciously selected for each case so that the responses have similar durations though each is sufficiently long and exhibits the desired regimes. Angular motions are predicted through numerical integration\(^17\) of equation (1); several variable-step Runge-Kutta algorithms intended for low to high mathematically stiff systems are utilized,\(^17\) but negligible differences between the results are found. Additionally, the maximum allowable time step for integration and the time resolution of resulting signals are set equal to approximately 78.1 \( \mu \)s, which corresponds to a sampling frequency of 12.8 kHz. The predicted angular

![Figure 1. Conceptual illustration of the non-linear torsional system: (a) single degree of freedom model; (b) elastic torque \( \Phi(\theta) \); and (c) examples of near backlash (\( \alpha = 0.1 \)) and pre-load (\( \alpha = 10 \)) non-linearities examined in this article. Here, \( \theta \) is the angular displacement, \( J \) is the torsional inertia, \( h \) is a Coulomb friction element, \( k(\theta) \) is a non-linear torsional stiffness element, and \( T(t) \) is the external torque. Refer to the text or Appendix 1 for all symbols.](image)
motions are shown in Figures 2 and 3; time and angular displacement are normalized by $\tau_{\text{off}}$ and $\Theta_i$, respectively. Responses of Figures 2 and 3 exhibit three distinct regimes, similar to prior work$^8$: (1) the double-sided impact regime (di) is characterized by $\theta$ crossing both $-\Theta_i$ and $\Theta_i$; (2) the single-sided impact regime (si) occurs when $\theta$ crosses $-\Theta_i$ only; and (3) the no-impact regime (ni) is defined when $\theta$ does not cross any $\Theta_i$.

The responses are further characterized by oscillatory period $\tau^{(j)}$ with corresponding time of occurrence $t^{(j)}$. Values $\tau^{(j)}$ and $t^{(j)}$ are defined as the elapsed and median times, respectively, between the $j^{\text{th}}$ and $(j + 2)^{\text{th}}$ extrema (peak or valley) of a signal, as shown in Figures 2 and 3. A continuous time domain signal $\tau(t)$ is then calculated from a smoothened spline fit$^17$ of the discrete points $(\tau^{(j)}, t^{(j)})$ where $j = 1, 2, \ldots, P$ and $P$ is the total number of oscillatory periods; calculated $\tau(t)$ is shown in Figure 4 for $\theta(t)$, $\Theta(t)$, and $\hat{\theta}(t)$. It is observed that $\tau(t)$ calculations are similar across all angular motions and exhibit regime-dependent asymptotic trends, summarized in Table 1. These trends can be described as hardening ($\tau(t) \to 0$), softening ($\tau(t) \to \infty$), or linear ($\tau(t) \approx \tau_{\text{off}}$) in nature. Next, a concept of instantaneous effective stiffness is proposed and later utilized to estimate $\tau(t)$.

### Concept of instantaneous effective stiffness

Assume that over a small interval of time, with period $\tau_e$ centered at time $t = t'$, the following approximation$^{14-16}$ can be made $\Phi(t)|_{t_e,t'} \approx \Phi(t)|_{t_e}$; here, $\Phi(t)|_{t_e}$ is a linear time-invariant approximation of $\Phi(t)|_{t_e,t'}$. Like prior work$^{15,16}$, $\Phi(t)|_{t_e}$ is defined by the following where $\hat{k}_m|_{t_e}$ and $\hat{k}_a|_{t_e}$ are mean and alternating stiffness components, respectively, and $\langle \theta(t), t' \rangle$, is the instantaneous expected value (windowed time average) of $\hat{\theta}(t)$ over period $\tau_e$ centered at time $t = t'$

$$\left. \Phi(t) \right|_{t_e,t'} = \left. \hat{k}_m(t, t') + \hat{k}_a(\theta(t) - \langle \theta(t), t' \rangle) \right|_{t_e}$$  \hspace{1cm} (5)

The proposed instantaneous expected value operator is defined by the following equation where $w_{\text{shp}}(t - t')$ is a windowing function

$$\langle \theta, t' \rangle = \frac{\int t w_{\text{shp}}(t - t')d\tau}{\int w_{\text{shp}}(t - t')d\tau}$$  \hspace{1cm} (6)

The windowing function $w_{\text{shp}}(t - t')$ has the following general form, chosen such that the instantaneous expected value operator can truly be localized.
Figure 3. Predicted angular motions (normalized) for a pre-load type non-linear system ($\alpha = 10$). Here, $\ddot{\theta}$ is the angular displacement, $\dot{\theta}$ is the angular velocity, $\theta$ is the angular acceleration, $T^f$ is oscillatory period, $t_{di}$ is the time of transition from the double-sided to single-sided impact regimes, $t_{si}$ is the time of transition from the single-sided to no-impact regimes, and $\dot{\theta}_1$ is the angular transition from stage I to II. Key: (---) angular motion; (---) $T^f$; and (__) $t_{di}$, $t_{si}$.

Figure 4. Calculated oscillatory period $T^f(t)$ (normalized) for a torsional system with: (a) backlash type non-linearity ($\alpha = 0.1$) and (b) pre-load type non-linearity ($\alpha = 10$). Here, $T^*_{II}$ is the natural period of stage II, $t_{di}$ is the time of transition from double-sided to single-sided impact regime, and $t_{si}$ is the time of transition from the single-sided to no-impact regime. Key: (---) calculated from $\ddot{\theta}(t)$; (---) calculated from $\dot{\theta}(t)$; (---) calculated from $\dot{\theta}(t)$; (---) $T^*_{II}$; and (__) $t_{di}$, $t_{si}$. 
in time

\[
\begin{align*}
\dot{w}_{\text{shp}}(t - \tau) = & \begin{cases} 0 & t < t^* - 0.5\tau_w \\
> 0 & t^* - 0.5\tau_w \leq t \leq t^* + 0.5\tau_w \\
= 0 & t > t^* + 0.5\tau_w \end{cases} \\
\end{align*}
\]

To estimate \( \hat{k}_{m,\text{ie},f} \) and \( \hat{k}_{a,\text{ie},f} \), the following error is defined:\end{itemize}

\[
e^2_{\text{ie},f} = \{\Phi(\theta) - \hat{\Phi}(\theta)\}^2_{\text{ie},f} 
\]

Next, \( \{e^2_{\text{ie},f} \}_t^{-1} \) is minimized with respect to \( \hat{k}_{m,\text{ie},f} \) and \( \hat{k}_{a,\text{ie},f} \) as follows

\[
\frac{\partial(e^2_{\text{ie},f})}{\partial \hat{k}_{m,\text{ie},f}} = 0, \quad \frac{\partial(e^2_{\text{ie},f})}{\partial \hat{k}_{a,\text{ie},f}} = 0
\]

Expanding equation (9), \( \hat{k}_{m,\text{ie},f} \) and \( \hat{k}_{a,\text{ie},f} \) are defined by

\[
\hat{k}_{m,\text{ie},f} = \frac{\langle \Phi(\theta), \langle \theta, \theta \rangle \rangle}{\langle \theta, \theta \rangle^2} \\
\hat{k}_{a,\text{ie},f} = \frac{\langle \Phi(\theta), \langle \theta, \theta \rangle \rangle - \langle \Phi(\theta), \langle \theta, \theta \rangle \rangle}{\langle \theta, \theta \rangle^2}
\]

Following prior work,\end{itemize}

\[
\hat{k}_{\text{ie},f} = \hat{k}_{a,\text{ie},f}
\]

However, \( \hat{k}_{\text{ie},f} \) becomes undefined under static conditions: \( \tau_f = [t_f + 0.5\tau_w, t_f - 0.5\tau_w] \) where \( t_0 \) and \( t_f \) are the initial and final times of the dynamic response, respectively. Furthermore, \( \tau_f \) and \( \tau_w \) are restricted by sampling parameters and the duration of the dynamic response. Finally, a continuous time domain signal \( \hat{k}(t) \) is calculated from a smoothened spline fit \( \hat{k}_{s}^{\text{fit}} \) of the discrete signal \( \hat{k}_{\text{ie},f}^{\text{dis}} \).

For the sake of illustration, \( \hat{k}(t) \) (normalized, \( \hat{k}_{\text{II}} = 1 \)) is estimated over several \( w_{\text{shp}} \) and a range \( \tau_{w} \) for a near backlash type non-linearity (\( \alpha = 0.1 \)). The normalized window length \( \tau_w \) is limited to \( \{0.5, 1, 2\} \), and windowing function \( w_{\text{shp}} \) is simply defined by the following (illustrated in Figure 5): (1) box-car \( w_{\text{box}} \); (2) triangular \( w_{\tri} \); and (3) right-skewed saw-tooth \( w_{\text{saaw}} \). These three window shapes are chosen so that comparisons can be made between uniform and non-uniform weighting, and symmetric and asymmetric windows. The effect of window length \( \tau_w \) is shown in Figure 6 where \( w_{\text{box}} \) is utilized. It is clear that a smaller value of \( \tau_w \) produces a local estimation of \( \hat{k}(t) \); likewise, a larger value produces a global or smoothened estimation. Additionally, it is observed that the range of \( \hat{k} \) is bounded by physical stiffness elements \( \hat{k}_{I} \) and \( \hat{k}_{\text{II}} \). The effect of window shape is shown in Figure 7 where \( \tau_w = 1 \).

As before, \( \hat{k}(t) \) is bounded by \( \hat{k}_{I} \) and \( \hat{k}_{\text{II}} \) for all window shapes; however, each produces a unique result. Windows shapes which have a uniform weighting \( (w_{\text{box}}) \) give a smoothened estimation of \( \hat{k} \) compared to the windows that do not \( (w_{\tri} \) and \( w_{\text{saaw}} \). It is also observed that estimations from asymmetric windows \( (w_{\text{saaw}}) \) exhibit a time-shift effect when compared to the symmetric windows \( (w_{\text{box}} \) and \( w_{\tri} \).

It is important to note that \( \hat{k}(t) \) is a mathematical concept (and not a physical spring element). Additionally, \( \hat{k}_{\text{ie},f} \) is significantly different from \( \langle \langle \hat{k}(\theta), \theta \rangle \rangle \rangle \), which is the instantaneous expected value of \( k(\theta) \) at \( t = t^* \) over \( \tau_w \). This is simply illustrated by the comparison of \( \hat{k}(\theta) \) to \( \hat{k}_{\text{ie},f}(\theta) \) (a continuous time domain signal of \( \langle \langle \hat{k}(\theta), \theta \rangle \rangle \rangle \)) for \( \tau_w = 1 \) and \( w_{\text{box}} \), as shown in Figure 8; the greatest absolute difference of \( \hat{k}_{\text{ie},f}(\theta) \) with respect to \( \hat{k}(\theta) \) roughly 35%.

![Figure 5. Uniform windowing functions](image-url)
Figure 6. Instantaneous effective stiffness $\tilde{k}(t)$ (normalized) for a backlash type non-linear system ($\alpha = 0.1$) with windowing function $w_{\text{box}}$ and uniform window lengths of: (a) $T_w = 0.5$; (b) $T_w = 1$; and (c) $T_w = 2$. Here, $\tilde{z}(1)$ and $\tilde{z}(P)$ are the domain limits of $\tilde{z}(t)$, $\tilde{z}_a$ and $\tilde{z}_b$ are regime transition times, and $k_I$ and $k_{II}$ are the torsional stiffness values for stages I and II, respectively. Key: (—)$\tilde{k}(t)$; (–)$\tilde{z}(1)$, $\tilde{z}(P)$, $\tilde{z}_a$, $\tilde{z}_b$; and (–––)$k_I$, $k_{II}$.

Figure 7. Instantaneous effective stiffness $\tilde{k}(t)$ (normalized) for a backlash type non-linear system ($\alpha = 0.1$) with uniform window length $T_w = 1$ and the following windowing functions: (a) $w_{\text{box}}$; (b) $w_{\text{tri}}$; and (c) $w_{\text{saw}}$. Here, $\tilde{z}(1)$ and $\tilde{z}(P)$ are the domain limits of $\tilde{z}(t)$, $\tilde{z}_a$ and $\tilde{z}_b$ are regime transition times, and $k_I$ and $k_{II}$ are the torsional stiffness values for stages I and II, respectively. Key: (—)$\tilde{k}(t)$; (–)$\tilde{z}(1)$, $\tilde{z}(P)$, $\tilde{z}_a$, $\tilde{z}_b$; and (–––)$k_I$, $k_{II}$.
Instantaneous effective \( \tilde{k}(t) \) and mean \( \bar{k}_m(t) \) stiffness values (normalized) for a backlash type non-linear system (\( \alpha = 0.1 \)) with uniform window length \( \tau_w = 1 \), and windowing function \( w_{exp} \). Here, \( \tilde{t}^{(1)} \) and \( \bar{t}^{(p)} \) are the domain limits of \( \tilde{t}(t) \), \( \tau_m \) and \( \tilde{t}_m \) are regime transition times, and \( \bar{k}_m \) is the torsional stiffness of stage II. Key: ( )—\( \tilde{k}(t) \); ( )—\( \bar{k}_m(t) \); ( )—\( \tilde{t}(t) \), \( \bar{t}^{(p)} \), \( \tau_w \); and (—)—\( \bar{k}_{ad} \).

Effective undamped linear time-invariant system with instantaneous effective stiffness \( \tilde{k}_{lin} \) at time \( t = \tau \) where \( \theta \) is the angular displacement and \( J \) is the torsional inertia.

Equivalent linear system with instantaneous stiffness

At time \( t = t' \), the SDOF non-linear system of Figure 1 may be approximated by an undamped time-invariant linear system (conceptually illustrated in Figure 9) consisting of torsional inertia \( J \) and torsional stiffness \( \tilde{k}_{lin} \). The corresponding natural period (linear system) is

\[
\tilde{\tau}_n|_{lin} = 2\pi \left( J/\tilde{k}_{lin} \right)^{0.5}
\]

A continuous time domain signal \( \tilde{\tau}_n(t) \) is calculated from a smoothened spline fit of \( \tilde{\tau}_n|_{lin} \); it is assumed that \( \tilde{\tau}(t) = \tilde{\tau}_n(t) \) where \( \tilde{\tau}(t) \) is the estimated oscillatory period of the non-linear system. The normalized signal \( \tilde{\tau}(t) \) is estimated for \( \tilde{\tau}_n \in [0.1, 6] \) across all \( w_{sup} \); \( \tilde{\tau}(t) \) for a near backlash type non-linearity (\( \alpha = 0.1 \)) and \( \tilde{\tau}_n = 2 \) is shown in Figure 10. For the sake of comparison to \( \tilde{\tau}(t) \) (which is calculated from \( \theta(t) \)), the domain of \( \tilde{\tau}(t) \) is limited to \( \tilde{t} \in [\tilde{t}^{(1)}, \tilde{t}^{(p)}] \). The quantitative agreements between calculated \( \tilde{\tau}(t) \) and estimated \( \tilde{\tau}(t) \) in the double-sided impact regime (di), single-sided impact regime (si), and the overall response (ov) are defined by metrics \( \Pi_{\tau} \) where \( \tau = \{di, si, ov\} \)

\[
\Pi_{di} = \exp \left[ -\int_{\tilde{t}^{(1)}}^{\tilde{t}^{(p)}} \left| \tilde{\tau}(t) - \tilde{\tau}(t) \right| dt \right]
\]

\[
\Pi_{si} = \exp \left[ -\int_{\tilde{t}^{(1)}}^{\tilde{t}^{(p)}} \left| \tilde{\tau}(t) - \tilde{\tau}(t) \right| dt \right]
\]

\[
\Pi_{ov} = \exp \left[ -\int_{\tilde{t}^{(1)}}^{\tilde{t}^{(p)}} \left| \tilde{\tau}(t) - \tilde{\tau}(t) \right| dt \right]
\]

If \( \Pi_{\tau} = 1 \), then \( \tilde{\tau}(t) = \tilde{\tau}(t) \) within the corresponding regime (deemed “best agreement”); otherwise, as \( \Pi_{\tau} \to 0 \), then \( |\tilde{\tau}(t) - \tilde{\tau}(t)| \to \infty \) (possibly the “worst agreement”). Metrics \( \Pi_{\tau} \) for all \( w_{sup} \) and \( \tilde{\tau}_w \) are shown in Figure 11 (near backlash type non-linear, \( \alpha = 0.1 \)) where \( R[\tilde{\tau}_w] \) is the range of \( \tilde{\tau}(t) \) in the corresponding response regime. It is easily noted that all window shapes exhibit similar trends; there is increasing accuracy from \( \tilde{\tau}_w = 0.1 \) to 1, followed by a leveling-off for roughly \( 1 \leq \tilde{\tau}_w < 4 \), and decreasing accuracy for \( \tilde{\tau}_w > 4 \). However, the box-car window \( w_{exp} \) has a slight loss of accuracy within \( 1 \leq \tilde{\tau}_w \leq 2 \). The maximum values of \( \Pi_{\tau} \) and corresponding \( \tilde{\tau}_w \), say \( \Pi_{\tau,\text{best}} \) and \( \tilde{\tau}_{\tau,\text{best}} \), are listed in Tables 2 and 3, respectively. It is obvious that no single pairing of \( w_{exp} \) and \( \tilde{\tau}_w \) (within this limited set) achieves best accuracy for all response regimes and given range of \( \alpha \) values, and that \( \tilde{\tau}_{\tau,\text{best}} \) has a wide range across response regimes.

Nonetheless, one important observation is made: Best \( \Pi_{\tau} \) is achieved when \( R[\tilde{\tau}_w] \leq \tilde{\tau}_w \leq 3 R[\tilde{\tau}_w] \). However, a uniform \( \tilde{\tau}_w \) might not satisfy this condition across all response regimes. Therefore, it stands to reason that a windowing function \( w_{sup} \) that utilizes an adaptive window length \( \tau_w(t') \) could improve the accuracy. Therefore, the following new adaptive method is proposed: (1) An initial window length \( \tau_w \) is judiciously chosen; (2) Estimation of \( \tilde{k}_{lin} \) is first estimated at time \( t' = 0.5 \tau_w \) using the uniform
Figure 10. Calculated $\bar{\tau}(\bar{t})$ and estimated $\tilde{\tau}(\bar{t})$ oscillatory periods (normalized) for a backlash type non-linear system ($\alpha = 0.1$) with uniform window length $\tau_w = 2$ and the following windowing functions: (a) $w_{\mathrm{box}}$; (b) $w_{\mathrm{tri}}$; and (c) $w_{\mathrm{aw}}$. Here, $\bar{\tau}^{(1)}$ and $\bar{\tau}^{(p)}$ are the domain limits of $\bar{\tau}(\bar{t})$ and $\bar{t}_d$ and $\bar{t}_u$ are regime transition times. Key: (---) $\tau^{(1)}$; (---) $\tau^{(p)}$; (---) $\bar{\tau}(\bar{t})$; and (---) $\bar{\tau}(\bar{t})$. 

Figure 11. Regime-dependent metrics $\Pi_{\mathrm{rr}}$ for a backlash type non-linear system ($\alpha = 0.1$) with uniform window $w_{\mathrm{aw}}$. Here, $\tau_w$ is the uniform window length (normalized), “di” is the double-sided impact regime, “si” is the single-sided impact regime, “ov” is the entire response, and $R[\bar{\tau}_{\mathrm{rv}}]$ is the range of $\bar{\tau}(\bar{t})$ (observed) in regime “rr”. Key: (---) $w_{\mathrm{box}}$; (---) $w_{\mathrm{tri}}$; (---) $w_{\mathrm{aw}}$; and (---) $R[\bar{\tau}_{\mathrm{rv}}]$. 

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Here, \( \tau_n \) is the time immediately prior to \( t' \), \( \beta \) is an arbitrary window length factor \((\beta > 0)\), and \( \tau_n|_{t'=t^-} \) is the natural period of the time-invariant linear system at \( t = t^- \):

\[
\tau_n(t') = 2\beta(\hat{k}/\tilde{f})^{0.5} = \beta\tau_n|_{t=t^-^+}.
\]  

(17) and (18)

and (4) Estimated oscillatory period \( \hat{f}|_{t=t^-} \) (non-linear system) is then calculated from \( \hat{k}|_{t=t^-} \), and the related continuous time domain signal \( \tilde{t}(t) \) is calculated. 17

For example, \( \tilde{t}(t) \) is estimated using \( w_{shp \_adp} \) over \( 0.1 \leq \beta \leq 6 \), and the corresponding metrics \( \Pi_{rr} \) are shown in Figure 12 for a near backlash type non-linearity \((\alpha = 0.1)\); here, \( \tau_{rr} = 2\beta(1) \).

Accuracy increases from \( \beta = 0.1 \) to 1, then reaches its maximum within \( 1 \leq \beta \leq 4 \), and decreases for \( \beta > 4 \). The \( \beta \) values for \( \Pi_{rr \_best} \) (say \( \beta_{rr \_best} \)) are given in Table 4. Similar to \( \tau_{rr \_best} \) for the uniform window, \( \beta_{rr \_best} \) varies greatly across the response regimes; however, \( 1 \leq \beta_{ov \_best} \leq 3 \) (roughly), which is consistent with the prior observation concerning \( \tau_n \). While \( \beta_{ov \_best} \) is similar across \( \alpha \) for \( w_{box \_adp} \) and \( w_{tri \_adp} \), it has significant variation for \( w_{saw \_adp} \).

A comparison of \( \Pi_{ov \_best} \) for the

<table>
<thead>
<tr>
<th>Window shape (shp)</th>
<th>( \Pi_{shp _best} )</th>
<th>( \Pi_{tri _best} )</th>
<th>( \Pi_{ov _best} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>(a) ( \alpha = 0.1 )</td>
<td>Box-car (box)</td>
<td>0.91</td>
<td>0.86</td>
</tr>
<tr>
<td></td>
<td>Triangular (tri)</td>
<td>0.91</td>
<td>0.89</td>
</tr>
<tr>
<td></td>
<td>Saw-tooth (saw)</td>
<td>0.92</td>
<td>0.82</td>
</tr>
<tr>
<td>(b) ( \alpha = 10 )</td>
<td>Box-car (box)</td>
<td>0.84</td>
<td>0.90</td>
</tr>
<tr>
<td></td>
<td>Triangular (tri)</td>
<td>0.83</td>
<td>0.91</td>
</tr>
<tr>
<td></td>
<td>Saw-tooth (saw)</td>
<td>0.83</td>
<td>0.92</td>
</tr>
</tbody>
</table>

Table 3. Normalized window length \( \tilde{t}_{rr \_best} \) necessary for \( \Pi_{rr \_best} \) using uniform window \( w_{shp \_adp} \).

<table>
<thead>
<tr>
<th>Window shape (shp)</th>
<th>( \tilde{t}_{rr _best} )</th>
<th>( \tilde{t}_{tri _best} )</th>
<th>( \tilde{t}_{ov _best} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>(a) ( \alpha = 0.1 )</td>
<td>Box-car (box)</td>
<td>1.1</td>
<td>2.5</td>
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<tr>
<td></td>
<td>Triangular (tri)</td>
<td>2.0</td>
<td>4.1</td>
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<td></td>
<td>Saw-tooth (saw)</td>
<td>5.4</td>
<td>1.7</td>
</tr>
<tr>
<td>(b) ( \alpha = 10 )</td>
<td>Box-car (box)</td>
<td>1.7</td>
<td>3.1</td>
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<td></td>
<td>Triangular (tri)</td>
<td>2.4</td>
<td>1.9</td>
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<tr>
<td></td>
<td>Saw-tooth (saw)</td>
<td>6.0</td>
<td>3.5</td>
</tr>
</tbody>
</table>

Figure 12. Regime-dependent metrics \( \Pi_{rr} \) for a backlash type non-linear system \((\alpha = 0.1)\) with adaptive window \( w_{shp \_adp} \). Here, \( \beta \) is the window length factor, “di” is the double-sided impact regime, “si” is the single-sided impact regime, “ov” is the entire response, and \( R[\tilde{t}_{rr}] \) is the range of \( \tilde{t}(t) \) (observed) in regime “rr.” Key: (---)----\( w_{box \_adp} \); (----)----\( w_{tri \_adp} \); (-----)----\( w_{saw \_adp} \); and (())----\( R[\tilde{t}_{rr}] \).
uniform and adaptive windows, $\Pi_{\text{ov, best}}(w_{\text{shp}})$ and $\Pi_{\text{ov, best}}(w_{\text{shp, adp}})$, respectively) is shown in Table 5; $w_{\text{shp, adp}}$ improves overall accuracy for most but not all cases of $\alpha$ and window shape. Given these limited observations, it is recommended that $w_{\text{shp, adp}}$ with $\beta \approx 2$ be utilized to achieve best possible accuracy in $\hat{\tau}(t)$ estimations. Although it is not addressed here, proper selection of $\tau_{\text{ov}}$ is also significant. Furthermore, $\tau_d(t')$, and thus $\beta$, are restricted by a choice of sampling parameters and the length of the dynamic response, similar to $t'$ and $\tau_d$.

In summary, the following observations are made regarding the window process: (1) Amplitude of $\hat{k}(t)$ is bounded by the stiffness values of the corresponding real spring for all windowing parameters; (2) Smaller window lengths produce a more temporally localized value, unlike larger lengths which give a more global value; (3) Uniformly and non-uniformly weighted window shapes (e.g. box-car $w_{\text{box}}$ and triangular window $w_{\text{tri}}$ shapes, respectively) of the same length can produce significantly different amplitudes; and (4) Asymmetric window shapes (e.g. saw-tooth $w_{\text{saw}}$) can produce a leading or lagging effect with respect to symmetric window shapes (e.g. box-car and triangular) of similar length.

### Validation of estimation method using measurements from two laboratory experiments

To validate the proposed method, two laboratory experiments (X1 and X2), which can be approximated by the SDOF non-linear model illustrated in Figure 1, are considered next. X1 is a controlled benchtop setup containing a single torsional clearance and X2 is a large-scale system containing a production vehicle clutch damper (with four stages). The elastic function $\Phi(\theta)$ for each experiment is shown in Figure 13. External torque $T(t)$ is applied in a step-like manner with initial and final operating points lying on separate stages, as shown in Figure 13. The primary operating stages of X1 and X2 are II and III, respectively. Thus, the experiments are sized such that $\tau_{\text{II, f}} \approx 0.1 \text{ s}$ in X1 and $\tau_{\text{II, f}} \approx 0.1 \text{ s}$; furthermore, normalized time for each system is scaled by these values. The time-varying oscillatory period $\tau(t)$ is calculated from measurements, as described in the section “Time-varying oscillatory period,” and shown in Figures 14 and 15; angular motions $\theta$ are omitted for the sake of brevity. For experiment X1, only the double-sided impact regime (di) with a softening trend occurs; however, experiment X2 demonstrates all possible regimes and trends (although regime transition times are not labeled). This difference can be attributed to the following factors: (a) In X1, the final operating point $(\theta_f, T_f)$ is $(0, 0)$, whereas in X2, $(\theta_f, T_f) \approx (\theta_\text{i}, > 0)$, which increases the likelihood that both double (di) and single-sided (si) impact regimes occur and (b) Energy dissipation has a large contribution to the response of X1 due to the

### Table 4. Window length coefficient $\beta_{\text{ov, best}}$ necessary for $\Pi_{\text{ov, best}}$ using adaptive window $w_{\text{shp, adp}}$

<table>
<thead>
<tr>
<th>Window shape (shp)</th>
<th>$\beta_{\text{di, best}}$</th>
<th>$\beta_{\text{sh, best}}$</th>
<th>$\beta_{\text{ov, best}}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>(a) $\alpha = 0.1$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Box-car (box)</td>
<td>1.0</td>
<td>3.3</td>
<td>1.0</td>
</tr>
<tr>
<td>Triangular (tri)</td>
<td>1.7</td>
<td>3.7</td>
<td>2.1</td>
</tr>
<tr>
<td>Saw-tooth (saw)</td>
<td>4.3</td>
<td>1.4</td>
<td>2.3</td>
</tr>
<tr>
<td>(b) $\alpha = 10$</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Box-car (box)</td>
<td>1.1</td>
<td>2.4</td>
<td>1.1</td>
</tr>
<tr>
<td>Triangular (tri)</td>
<td>2.8</td>
<td>2.0</td>
<td>2.0</td>
</tr>
<tr>
<td>Saw-tooth (saw)</td>
<td>4.6</td>
<td>3.5</td>
<td>3.5</td>
</tr>
</tbody>
</table>

### Table 5. Summary of $\Pi_{\text{ov, best}}$ for uniform and adaptive windows, $\Pi_{\text{ov, best}}(w_{\text{shp}})$ and $\Pi_{\text{ov, best}}(w_{\text{shp, adp}})$, respectively.

<table>
<thead>
<tr>
<th>Window shape (shp)</th>
<th>$\Pi_{\text{ov, best}}(w_{\text{shp}})$</th>
<th>$\Pi_{\text{ov, best}}(w_{\text{shp, adp}})$</th>
</tr>
</thead>
<tbody>
<tr>
<td>(a) $\alpha = 0.1$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Box-car (box)</td>
<td>0.778</td>
<td>0.779</td>
</tr>
<tr>
<td>Triangular (tri)</td>
<td>0.809</td>
<td>0.810</td>
</tr>
<tr>
<td>Saw-tooth (saw)</td>
<td>0.711</td>
<td></td>
</tr>
<tr>
<td>(b) $\alpha = 10$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Box-car (box)</td>
<td>0.736</td>
<td>0.742</td>
</tr>
<tr>
<td>Triangular (tri)</td>
<td>0.750</td>
<td>0.749</td>
</tr>
<tr>
<td>Saw-tooth (saw)</td>
<td>0.755</td>
<td>0.747</td>
</tr>
</tbody>
</table>

**Figure 13. Illustration of multi-staged elastic torque curves $\Phi(\theta)$, and initial $(\theta_1, T_1)$ and final $(\theta_f, T_f)$ operating points for two experiments used to validate estimation methods: (a) experiment X1 and (b) experiment X2. Here, $\theta$ is the angular displacement, $T$ is external torque, $\theta_1$ are angular stage transitions, and subscripts (I, II, ...) denote stages.**
presence of a backlash non-linearity \((k_1 = 0)\), unlike X2 for which \(k_1\) is a very soft spring.

Following the suggestions from Section 5, \(\hat{r}(t)\) signals for both X1 and X2 are estimated using \(w_{\text{tri}_\text{adp}}\) with \(\beta = 2\); an initial guess of \(\hat{\tau}_{\text{w0}} = 1\) is selected. Quantitative comparisons between \(\hat{r}(t)\) (observed) and \(\hat{r}(t)\) (estimated) are shown in Figures 14 and 15 for experiments X1 and X2, respectively. In particular, the signal \(\hat{r}(t)\) correctly exhibits the hardening and softening nature observed in each experiment though it lacks complete agreement with \(\hat{r}(t)\). Several factors could contribute to this difference. First, the suggested windowing parameters and algorithm (developed from simplified example cases) may not be optimal for these real experiments. Furthermore, the corresponding non-linear models of these experiments may require additional degrees of freedom and the corresponding features would need a higher level of characterization. Finally, if significant viscous damping is present with a relatively low stiffness, then the oscillatory period cannot be accurately estimated by an effective natural period. Instead, an effective damped natural period would need to be considered.

**Conclusion**

The main contribution of the paper is the development of a new analysis tool that estimates the trends in time-varying oscillatory periods in the step-response of a torsional system containing a multi-staged spring. Development of the method begins with the formulation of a new concept of instantaneous effective stiffness. Although it is strictly mathematical in nature (and not a physical element), its amplitude is limited by the stiffness values of the corresponding real spring. The related computations
employ windowing functions, and thus a range of window shapes and lengths are investigated. Next, the instantaneous effective stiffness is used to approximate the real non-linear system (at some instant during its step-response) as an undamped time-invariant linear oscillator. Here, it is assumed that the damping in the real system is relatively low such that its oscillatory period at that time can be approximated by the natural period of the linear oscillator. The method is then applied to numerical example cases and a set of windowing parameters, which include window shape and an adaptive length algorithm, are suggested from the results. Finally, the proposed method is validated by correctly estimating the asymptotic trends (such as hardening or softening) observed in the time-varying oscillatory periods of two recently proposed experiments by the authors.\textsuperscript{8,17} Although measured (or numerically simulated) responses must be known to calculate the instantaneous effective stiffness, the utility of this concept demonstrates its relationship to the non-linear response. Specifically, this time domain concept correctly identifies hardening and softening trends, which have been historically discussed in the context of harmonic excitation.\textsuperscript{9,11,16} The method of this article should serve as an important diagnostic tool for the system identification of an unknown device. Assuming that angular displacement and torque can be measured for a step-response, the proposed method should provide value insight for the characterization of non-linear features (say using a SDOF approximation), as well as amplitude dependence.\textsuperscript{19} In cases of relatively high viscous damping, the method could be applied to only the first few cycles of oscillations. Overall, the proposed method should improve the efficiency and accuracy of the model building process for mechanical devices with clearance non-linearities.\textsuperscript{2,19}

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References


Appendix I

Notation

eerror

hCoulomb friction amplitude

Jtorsional inertia

ktorsional stiffness

Ntotal number of stiffness stages

ttime

Texternal torque

wwindowing function

αwindow length coefficient

βwindow length coefficient

Γtotal torque transmission

ηregularizing factor for Coulomb friction

θ, ⃗{θ}, ⃗{θ}angular displacement, velocity, and acceleration

Θstage transition (angular)

Πestimation metric

τperiod of oscillation

Φelastic torque

Ξunit-step function

Ψdissipative torque

di double-sided impact

f final point

mmean coefficient

nnatural

nin no-impact

oin initial point

ovoverall response

rupright

sawsaw-tooth

shpshape

ISISingle-sided impact

tinstantaneous (time dependent)

tritriangular

wwindowing function

I, II, ... stage index

Superscripts

(i) oscillatory period number

(i = 1, 2, ..., P)

~effective

Subscripts

a alternating coefficient

adpreadaptive window

best”best”

boxbox-car