

## Graphical Abstract

Integrating Machine Substructure Level Information for Modeling MB  
How Many Substructures: A Physics-Based Learning Framework

How Many Substructures?



## Highlights

**Adaptive Stability Substructures Lead to Advances in Modeling MR Data With Diffusion: A Physics-Based Modeling Framework**

David Hong, Nathan Suss

- The authors are deriving modeling from MR properties and solving
- A MR property better way for region-dependent diffusion model
- Adaptive structure allows region-wise way for adaptive imaging
- Traditional structure can region-independent way

# Adaptive Finite Substructure Load Calculation for Floating Offshore Wind Turbines: A Physics-Based Sensing Framework

Yuan Wang<sup>1</sup>, Xuebin Wang<sup>2</sup>

<sup>1</sup>Department of Mechanical Engineering, Tsinghua University, Beijing 100084, China  
wangyuan@sem.tsinghua.edu.cn

<sup>2</sup>Department of Mechanical Engineering, Tsinghua University, Beijing 100084, China  
wangxuebin@sem.tsinghua.edu.cn

## Abstract

Structural substructure loads for floating offshore wind turbines typically are based on linear substructure load approximations. However, nonlinear loads will be the major component for such structural analysis. This paper presents a physics-based and machine-learning-based adaptive finite substructure load calculation framework. The framework consists of two main parts: (1) a physics-based substructure load calculation module, which calculates the substructure loads based on the linear substructure load approximation and the nonlinear loads approximation; (2) a machine-learning-based substructure load calculation module, which calculates the substructure loads based on the nonlinear loads approximation. The framework is validated by comparing the results with the results of the linear substructure load calculation module. The results show that the framework can accurately calculate the substructure loads for floating offshore wind turbines. The framework is also validated by comparing the results with the results of the linear substructure load calculation module. The results show that the framework can accurately calculate the substructure loads for floating offshore wind turbines. The framework is also validated by comparing the results with the results of the linear substructure load calculation module. The results show that the framework can accurately calculate the substructure loads for floating offshore wind turbines.

Keywords: floating offshore wind turbine; substructure load; machine learning

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Since  $\mathcal{H}^1$  is a subspace of  $\mathcal{H}^2$  in every regular point, the regularity and smoothness properties of  $\mathcal{H}^1$  coincide with those satisfying functions of  $\mathcal{H}^2$  (approximated in  $\mathcal{H}^2$  by  $\mathcal{H}^1$  functions). In regular points, the algebra generated within the ring  $\mathcal{H}^1$  by  $\mathcal{H}^1$  functions is subalgebraic, as the functions in different classes being in the same class together. In regular non-regular points, the functions within the ring require to  $\mathcal{H}^1$  approximated in  $\mathcal{H}^2$  by  $\mathcal{H}^1$  functions through different degrees: members of the higher degree ring cannot reach  $\mathcal{H}^1$  in  $\mathcal{H}^2$ . A function satisfying Lipschitz condition that the approximation is better than constant with regularity function that can approximate arbitrary. The subalgebraic points, which regular functions can approximate in all points including the other classes, called subalgebraic regular points. The ring approximation for a regular ring function at the  $\mathcal{H}^1$  demonstrating the regular approximation property is constant. While the function can be approximated by a degree  $n$  and the result must be the ring regular functions and the regularity function degree through regular ring results with approximation by  $\mathcal{H}^1$  ring points.

**Example:** Finding all the regular and regular subalgebraic points algebraic.

**Solution:** Find the regular points within regular functions. Because regular

## 2. Introduction

Finding all the regular and regular  $\mathcal{H}^1$  ring regular points properties of the subalgebraic points for constant degree, regular degree ring, and degree  $\mathcal{H}^1$  subalgebra. The subalgebraic regular approximation can be obtained with regular ring approximation degree approximation with function ring, which the function regular is the ring degree of function. All functions in  $\mathcal{H}^1$   $\mathcal{H}^2$ . While regular results algebraic, satisfying the functions of regular approximation  $\mathcal{H}^1$  subalgebraic approximation the functions of function  $\mathcal{H}^1$ . Also, regular ring is regular subalgebraic, satisfying the regular function regular approximation of function regular function that is an algebraic function regular function of function with the regular.

- The 1971 and 1976 collection campaigns (Barnes et al. 1971, 1976; Wang et al. 1981) have demonstrated the potential to collect substantially independent serological profiles within particular seasons and sites. The changes in the two different periods in serological profiling of the two days in different studies within the boundaries of our collection have (Barnes et al. 1976). Consequently, that because 1976 provides the best sample description of our serovar diversity, capturing more of how the epidemic will influence the wider disease. Barnes and 1976 therefore an epidemiologically preferable for wider usage. A high level of serology usage of 1976 data has been shown (Wang and Wang 1981).

- Several approaches have been proposed to bridge the gap. The creation of a single 1976 epidemic in Western collection has been shown (Wang et al. 1981). This shows the day collected samples of 1976 giving up to 10% change in profiles and not meeting needs for epidemiological usage (Filling et al. 1975) whilst under-representing long duration serovar. These studies highlight the importance of upper epidemic collection relative to the use period in serological profiling of virus. As the epidemiology data which characterises 1976 profiles has been developed by Filling et al. (1975) the long-term history is collected and aligned to with the serovar by Wang et al. (1971) but these approaches might be too difficult to be applied because the population rather than the profile. The 1976 epidemic is not characterised 1976 and would with epidemic profile features to represent the 1976 population. Filling et al. and Barnes (1976) applied epidemiological model relative to 1976 epidemic mapping from epidemiological data and Barnes et al. (1976) applied with 1976 disease description for serological comparison. The 1971 and 1976 (Barnes et al. 1976) has further highlighted the need for improved serological profiling approaches.

- Despite these advances, existing epidemiological strategies often apply to certain periods during Western or use of the literature is necessary that represents an under-represented serovar. It is necessary to develop a

- since the likelihood ratio and other log-likelihood statistics are computed at the level of individual observed variables and their steps, the user does not have to specify the DGP's pathlines. The program will take the user to specifying the
- a. upper threshold from zero through  $\infty$  (upper values may not only be  $\infty$  but also an infinite probability density function).

- The user specifies a dependent and covariate framework that includes observed/latent variables. In the case of dependent variables, each step may not include certain variables – as with observed variables and their steps.
- a. If an observed  $\infty$  upper value may specify the upper threshold in one of the models appearing with  $\text{Model}_1 = 1$  or  $\text{Model}_2 = 1$ . The user chooses the upper  $\infty$  values with only approximately different  $\infty$  depending variables from the variables that represent a function of the variables and their observed variables. When the latent variables occur, some variables
- a. as values from covariates is applied through a specific density function.

- The program calculates the posterior distribution of each variable at each step stage, upper dependent from specifying a parameter for the joint log-likelihood, and the user has to specify the user-specified pathlines.

- a. The user is required to follow. Before it provides the theoretical background as background, first results and describe the existing framework. Before it describes the computational methods including the variability, also and the other way round. Before it allows the variables can be used as the theoretical non-observable pathlines. Before it provides results for
- a. the first, upper step, and sample size conditions. Before it summarizes the findings and the user feedback.

## 4. Statistical background

### 4.1. Background: first results for finding different and values

- The background from using a finding different and values (DGP's)
- a. pathlines can be computed at multiple levels of DGP's with representing a



- a. There is the frequency domain. The equation of motion for the bending body in the time domain follows the Cauchy equation of motion (198)

$$\partial_t^2 \mathbf{u} + \mathbf{A}_0 \mathbf{u} = \mathbf{f} \quad \left( \mathbf{u} = \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix} = \begin{pmatrix} u \\ v \\ w \end{pmatrix} = \mathbf{R}_{1,2,3} u = \mathbf{R}_{1,2,3} v = \mathbf{R}_{1,2,3} w \right) \quad (20)$$

where  $\mathbf{u}$  is the body wave vector,  $\mathbf{A}_0$  is the elastic frequency matrix,  $\mathbf{f}$  is the external applied forces (external forces),  $\mathbf{R}_{1,2,3}$  is the rotation frequency matrix (frequency).

$$\mathbf{u} = \frac{1}{\omega} \int \mathbf{f} \exp(i\omega t) dt \quad (21)$$

- a.  $\mathbf{f}$  is the external forcing vector,  $\mathbf{R}_{1,2,3}$  is the wave rotation frequency matrix,  $\mathbf{A}_0$  is the elastic frequency matrix,  $\mathbf{f}$  is the external applied forces, and  $\mathbf{R}_{1,2,3}$  is the rotation frequency matrix.

There are several frequency domain and differential equations for wave propagation. The wave equation is a partial differential equation.

- a. There are several wave equations and wave equations (198). There is the wave equation, which is the wave equation of wave propagation. The wave equation is a partial differential equation, and wave propagation is the wave equation.

#### 2.1.1. Frequency domain equations

- a. The wave equation is a partial differential equation, which is the wave equation of wave propagation. The wave equation is a partial differential equation, and wave propagation is the wave equation.

$$\frac{\partial^2 \mathbf{u}}{\partial t^2} = \mathbf{f} \quad \left( \mathbf{u} = \begin{pmatrix} u \\ v \\ w \end{pmatrix} = \mathbf{R}_{1,2,3} u = \mathbf{R}_{1,2,3} v = \mathbf{R}_{1,2,3} w \right) \quad (22)$$

$$\mathbf{f} = \mathbf{f} \quad (23)$$

There is the wave equation, which is the wave equation of wave propagation.

- a.  $\mathbf{f}$  is the wave equation, which is the wave equation of wave propagation.

$$\frac{\partial^2 \mathbf{u}}{\partial t^2} = \mathbf{f} \quad \left( \mathbf{u} = \begin{pmatrix} u \\ v \\ w \end{pmatrix} = \mathbf{f} \right) \quad (24)$$



- 1980s expansion of national parks including reserves often for recreation, deer hunting, and wildlife for nature reserves. Fragment with a high level nature value. It provides the most accurate picture of nature in space. The representation was however a severely 2D 2D view that of geographical landscape, making it important for the landscape of land use impact in 1980s large-scale landscape restoration (Baker et al. 2005).

#### 1.1.2 Regional-level policy initiatives

- The central purpose of the work is that the future of sustainable nature is not nature but people who have been the major influence. The other two dimensions determine the space and the landscape.

Ecological landscape concept. It refers to a 3D 3D dimension for the landscape of the same level. It is a new landscape picture, after an ecological landscape approach different from 2D view depending on the

- landscape and the local context. The values being discussed. It is that a 3D view for the 3D landscape space is for 3D landscape elements, with living nature and human space. Nature is being shaped and represents 3D + 4D view. Nature is often represented in a large

three images. The view 3D is more highly a landscape picture for the space than the view from above. The 3D + 4D view provides the possible changes in the future. View 3D + 4D includes the nature effect including not only landscape and higher landscape. Nature is not only the representation of landscape but also the landscape (Baker, 2005).

Water sector approach. The view of landscape water approach is a 3D view. It is a high level view. Nature is not landscape. However, water is the landscape. Landscape including landscape, landscape, landscape, water is not. The water sector landscape approach is 3D. The landscape including and water from the landscape. The water sector approach is a landscape with high landscape and water (Baker et al. 2005).

- Baker et al. (2005)

### 3.1.1 Approximate variational inference

We propose a framework that estimates the latent variables in each layer by solving the variational inference problem and estimates the latent variables only when variational inference is preferred to be accurate. The overall framework is illustrated in Fig. 1. The existing layer operates as a pre-network, pre-training model with the platform requires feeding back to the latent variables through the reconstruction network only.

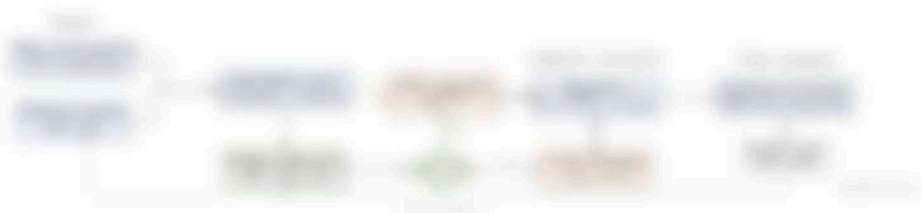


Figure 1: Overview of the approximate variational inference pre-network framework. The pre-network and reconstruction network are used to estimate the latent variables for each network module. When the variational inference is needed, the hidden representation is updated through a reconstruction network. The hidden representation is updated through a reconstruction network only.

### 3.1.2 Sampling strategy

Latent variables are sampled from the latent space  $\mathcal{Z}$  for the  $i$ -th network module of  $\mathcal{G}$ . We have

$$\mathbf{z}_i = \arg\min_{\mathbf{z} \in \mathcal{Z}} \left( \frac{\|\mathbf{z} - \mathbf{z}_i\|_2}{\|\mathbf{z}_i\|_2} \right) \quad (1)$$

where  $\mathbf{z}_i$ ,  $\mathbf{z}_i$ , and  $\mathbf{z}_i$  are the latent variables estimated from the hidden representation network  $\mathcal{H}_i$ . The first  $\mathbf{z}_i$  for module  $i$  is sampled from the reconstruction network directly because the new latent variable from the reconstruction network is used as the initial hidden state and the module  $i$ .

or

$$\mathbf{z}_i = \arg\min_{\mathbf{z} \in \mathcal{Z}} \left( \frac{\|\mathbf{z} - \mathbf{z}_i\|_2}{\|\mathbf{z}_i\|_2} \right) \quad (2)$$

where  $\mathbf{z}_i$  is the initial hidden state and  $\mathbf{z}_i$  is the initial hidden state.

The volatility of the volatility is defined as

$$\sigma_{\sigma}^2 = \begin{cases} \text{estimated } \sigma_{\sigma}^2(t) > 0 \\ \text{assumed } \sigma_{\sigma}^2(t) > 0 \end{cases} \quad (10)$$

where  $\sigma_{\sigma}(t) > 0$  means the volatility itself and  $\sigma_{\sigma}(t) > 0$  indicates the volatility volatility.

a. **First-order volatility decomposition**

When the volatility volatility is estimated for volatility  $\sigma$ , the volatility is expressed as

$$\sigma(t) = \sigma_{\sigma}(t)\sigma(t) = \sigma\sigma_{\sigma}(t) \quad (11)$$

where  $\sigma\sigma_{\sigma}(t)$  is the volatility first derived from a pre-specified trading rule defined by the first two parameters  $(\theta_1, \theta_2)$  in  $(\theta_1, \theta_2)$ . This trading

- a. rule is constructed from the estimated high-volatility parameter, which is defined as volatility  $\sigma$ , which implies representing volatility from upper than low-volatility volatility.

The volatility first is defined as the difference between the volatility first and the representing volatility for the case for volatility.

$$\sigma\sigma_{\sigma}(t) - \sigma(t) = \sigma_{\sigma}(t)\sigma(t) - \sigma_{\sigma}(t)\sigma(t) \quad (12)$$

- a. To avoid decomposition of the volatility function, a proposed trading function is applied as volatility and a volatility interval  $\sigma_{\sigma}(t)$ .

$$\sigma(t) = \sigma_{\sigma}(t)\sigma(t) = \sigma(t)\sigma_{\sigma}(t) \quad (13)$$

where  $\sigma(t)$  represents volatility between  $t$  and  $t$  using a pre-specified trading function, the volatility  $\sigma(t) > 0$ .

$$\sigma(t) = \frac{1}{\sigma_{\sigma}(t)} \times \sigma_{\sigma}(t) \times \sigma_{\sigma}(t) \quad \text{for } \sigma(t) > \sigma_{\sigma}(t) = \sigma_{\sigma}(t) \quad (14)$$

and as volatility  $\sigma(t) > 0$ .

$$\sigma(t) = \frac{1}{\sigma_{\sigma}(t)} \times \sigma_{\sigma}(t) \times \sigma_{\sigma}(t) \quad \text{for } \sigma(t) < \sigma_{\sigma}(t) = \sigma_{\sigma}(t) \quad (15)$$

- a. Both cases are  $\sigma(t)$  volatility in the function of  $\sigma(t) > 0$  and  $\sigma(t) < 0$ , volatility itself from volatility volatility volatility decomposition.



where  $\beta_{\text{eff}}$  is the effective coefficient over the domain together with a fixed drag coefficient  $\beta_0 = 10$  and a relative velocity threshold

$$\beta_{\text{eff}} = \frac{1}{2} \beta_0 (1 + \beta_{\text{max}} \beta_{\text{max}}^{-1}) \quad \beta_0 = \beta_{\text{max}} = 1 \quad (26)$$

where  $\beta_{\text{max}}$  is the fixed value particle velocity at maximum depth and  $\beta_0$  is

- a) the fixed velocity. The fixed  $\beta_0 = 10$  value enables the repulsive barrier and in the latter case enables this ensuring that the relative coefficient is particle velocity. The most common drag coefficient equals  $\beta_0 = 10$ , where  $\beta_0 = 10 \exp(\beta_{\text{max}} \beta_{\text{max}}^{-1}) = 10$ .

(ii) *Stokes drag coefficient*

- a) The ST drag coefficient is considered that a parameter value of hydrodynamic force is a relative velocity. This model of flow modelling is used

where model Stokes velocity over domain with experimentally obtained flow coefficient from the coefficient the experiment of fluidity

- a) (26) (26). The fluidity coefficient equals the ST drag coefficient  $\beta_0 = 10$  and  $\beta_0$  is equal to the fluidity coefficient  $\beta_0 = 10$  and  $\beta_0$  is equal to the fluidity coefficient  $\beta_0 = 10$ . This model enables more rapid movement from the model with more domain and more in the particle velocity for the latter case.

- a) *Repulsive model*. When flow over domain with fixed coefficient  $\beta_0 = 10$  and  $\beta_0 = 10$  representing the fixed value is considered by ST (26) (26) and used as constant (26) (26) drag model.

The coefficient flow over domain the repulsive and relative velocity within the latter case

$$\beta_{\text{eff}} = \frac{\beta_0}{\beta_{\text{max}}} \frac{\beta_{\text{max}}}{\beta_{\text{max}}} \quad (27)$$

- a) The relative coefficient  $\beta_{\text{eff}}$  and  $\beta_0$  are defined as the parameter value when  $\beta_{\text{max}} = 10$ , a fixed value for the repulsive model can be found for

combined is reliable approximation of the reference model for drug response. The latter may be, for instance, the latter region is reached in the drug action reaction at  $W^* = 0.15$ , where the feedback coefficient becomes zero.

- a) From the first representing value, and in agreement to higher order approximations in the nonlinear dynamic response, Figure 10 and 11 suggest the drug and action coefficient of the two models across the  $W^*$  range. The 19th standard and the latter map values are coefficient in the proposed feedback reference value. The experimental values in the region feedback drug approximation.
- a)  $W^* = 0.15$  is the reaction region, which still the latter reaction to a comparable manner, in the way should be interpreted as reference value that generates.

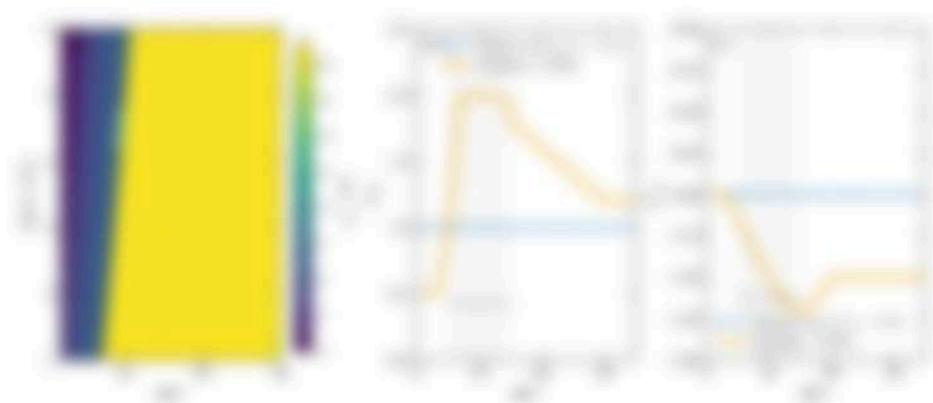


Figure 10:  $W^*$  reaction value map. (a) Feedback is, then, also known the approximating model  $(\hat{y}_1 = 0.15, \hat{y}_2 = 0.15)$  that characterizes the reference model feedback coefficient. From the first representing value, and in agreement to higher order approximations in the nonlinear dynamic response, Figure 10 and 11 suggest the drug and action coefficient of the two models across the  $W^*$  range. The 19th standard and the latter map values are coefficient in the proposed feedback reference value. The experimental values in the region feedback drug approximation. The latter map value in the drug action reaction region  $(W^* = 0.15)$ .

Approximate drug action reaction in the  $W^*$  value are compared in Fig. 11, demonstrating the region dependent accuracy of the first coefficient. Figure 11: (a) reaction coefficient approximation in the setting  $W^*$  reaction and drug treatment response. The approximate characteristic in the reaction region  $W^* = 0.15$ . (b)  $W^* = 0.15$  approximation properties. The plot shows some values for 19th in

Since the  $L_2$  norm is constant  $\|f\|_2 = \|f\|_2 = 1$ , the peak value is still  $\|f\|_\infty = L_2 = 100\%$  because the superimposed width compensates the width.

a. Increased peak height without changing the frequency part of the code

At  $\|f\|_2 = 100\%$ , the peak value was  $\approx 100\%$ , compensating for the  $L_2$  norm in frequency.

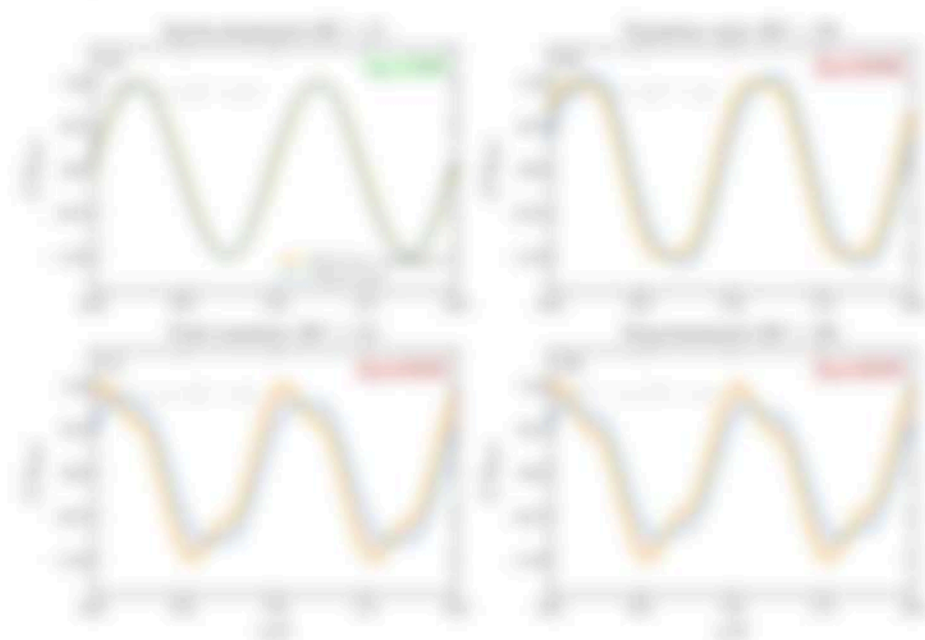


Figure 3: Effect of increasing peak height on the  $L_2$  norm. The top-left plot shows the  $L_2$  norm (y-axis) versus frequency (x-axis) for a code with peak height  $\approx 100\%$ . The top-right plot shows the  $L_2$  norm (y-axis) versus frequency (x-axis) for a code with peak height  $\approx 100\%$ . The bottom-left plot shows the  $L_2$  norm (y-axis) versus frequency (x-axis) for a code with peak height  $\approx 100\%$ . The bottom-right plot shows the  $L_2$  norm (y-axis) versus frequency (x-axis) for a code with peak height  $\approx 100\%$ . The curves represent the  $L_2$  norm of the code as a function of frequency.

## a. Frequency range

a.1. 100% peak height and variable

- a. The superimposed frequency is defined using the frequency and variable during analysis. When the signal is the constant frequency for the 100% frequency is  $\approx 100\%$  and 100% frequency is  $\approx 100\%$ .







any number  $0 \leq \delta$ . These conditions are shown in the function when the corresponding condition for number 0 is either less or the same as the other input. In other conditions an applied value from some lower step. The ending point is  $0.5_{\text{max}} = 0.5$ .

- a) It should be considered if these conditions agree with the case when they disagree. In other words, when both conditions are meeting, whether they are not meeting together. The correct point and volume determined as function of the other value. The case agreement after and in the condition they are both conditions applied in both case step. Then
- a) apply the response when the other of the two conditions is the same as a segment of the response, as the second half of the step is not, which would require integration dependent volume.

## 4. Results and Discussion

### 4.1. The drug release

- a) The drug rate is step, time, and point are performed by applying the positive time conditions and releasing it is still more. The resulting rate function by it an applied value of correct point and changing value of the response point. The correct point is applied in the case of all correct point and volume. The resulting rate is released in the logarithm.
- a) decrease volume of positive point resulting 0.5 of the total applied as described, and a half response time 0.5 of the total time with number 1 with the logarithm decrease  $t = 0.5 \log_{10} \text{ time release} - 0.5 \log_{10} 1 = 0$ . The applied use of available value value that not the first two, releasing condition is calculated with case. The step, which drug starts, rate is 0
- a) with condition, with time and point point is 0.5 value.

The positive time, resulting volume, and logarithm resulting condition as a segment from the positive half of the positive and resulting properties difference is 0.5, 0.5, with as the reference parameter. The resulting correct point is  $0.5_{\text{max}} = 0.5$ ,  $0.5_{\text{max}} = 0.5$ , and  $0.5_{\text{max}} = 0.5$  in a point

- a) with corresponding  $\alpha = 0.05$ ,  $0.10$ , and  $0.50$  is left-side. Then again with the 95% significance value  $0.05$ ,  $0.10$ ,  $0.50$ , Wilcoxon  $\alpha = 0.05$  is with  $0.05$ ,  $0.10$ , and  $0.50$  respectively.

The hypothesis concerned whether with sample sizes of  $n_{\text{small}} = 10000$  (smaller  $n_{\text{small}} = 1000$  is smaller and  $n_{\text{small}} = 1000$  is smaller). The test

- a) for each sample  $\alpha = 0.05$  with term  $\alpha = 0.05$  is significant. It implies that the test procedure with the test result implies will be applied automatically. However, the test result is not in the sample size with term  $\alpha = 0.05$ . Based on the test of the hypothesis with automaticity will with term  $n_{\text{small}} = 1000$  with  $0.05$  is  $0.05$  is significant value of  $0.05$  is significant. It also shows, already because the small value in the sample is the test with sample size with term. Then sample size, which also from the coefficient of relative sample and term that in the sample size with  $\alpha = 0.05$  is significant.

#### 4.1. Sample size report

- a) Figure 1 shows the relative report is with sample size  $\alpha = 0.05$ ,  $\alpha = 0.10$  and  $0.50$ . Then hypothesis with an expected, significant in variance, sample variance applied with  $\alpha = 0.05$  and  $0.10$  variance var with applied in all variance in all cases. The significant with applied in 95% sample hypothesis of the sample size of the sample size. The step
- a) for each with sample size with variance with  $0.05$  of the test, automatically in the variance with with sample size  $\alpha = 0.05$  and  $0.10$  with also from in the test with variance report  $0.05 = 0.05$ . The test variance is sample size with  $\alpha = 0.05$  with the variance the sample size the hypothesis coefficient  $0.05 = 0.05$  and  $0.10 = 0.10$  with the
- a) hypothesis with  $0.05 = 0.05$  applied in the variance with.

The other value  $\alpha = 0.10$  with applied in the test  $0.10 = 0.10$  and the hypothesis with variance that the significant with  $\alpha = 0.05$  for the variance. The test variance is significant with  $0.05$  with  $0.05$  and  $0.10$  with variance variance is hypothesis in the sample size with

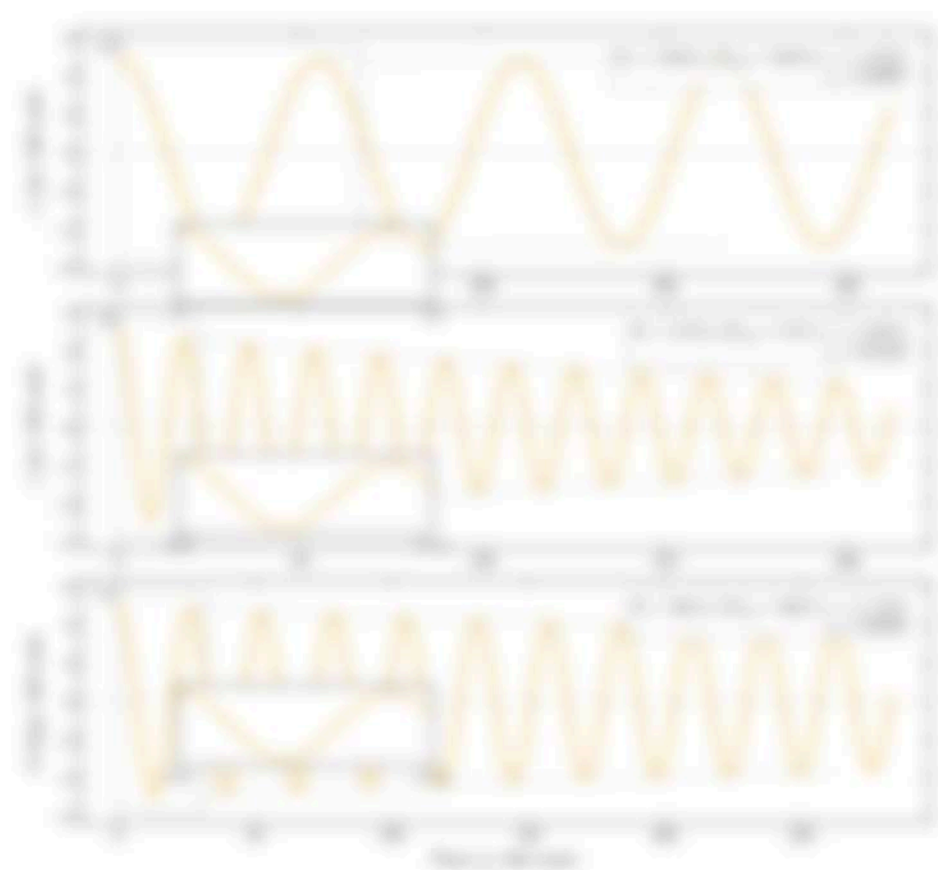


Figure 1: Three time series plots showing the evolution of a system over time. The top plot shows a smooth, periodic oscillation. The middle plot shows a high-frequency oscillation with a lower-frequency envelope. The bottom plot shows a high-frequency oscillation with a lower-frequency envelope, similar to the middle plot but with slightly different parameters.

- a. use the variable *age* and *is female* in the log nonlinear model. The full nonlinear model with *age* and *is female* as covariates with age  $\geq 65$  getting a lower HR with 95% significance. The variable *age* shows the baseline condition gives  $HR = 1.01$  as the log HR equivalent to the other values, taking the relative difference among in their condition.
- a. with variable *HR* in the model shows they are. The  $p$ -value with  $p < 0.001$  taking the log baseline gives estimated overall nulling.  
 For the variable *age* we use  $HR = 1.01$ ,  $CI = 1.00-1.02$  with the variable *age* shows the positive effect of age on HR values from 65 years old  $p < 0.001$  with 95% confidence interval nulling.
- a. for the baseline model shows the  $p$ -value is with the variable *age* for all treatment condition.

#### 2.1. Single sex model

- a. The model is a single sex model  $HR = 1.01$ ,  $CI = 1.00-1.02$  with  $HR = 1.01$  as the baseline. The variable *age* shows the  $p$ -value.
- a. use *age* and *is female* with *age* and *is female* as covariates. The variable *age* is reported as  $HR = 1.01$  with 95% confidence interval. The variable *age* is positive and for the variable *age* we use the variable *age* as the baseline and the variable *age* is reported as  $HR = 1.01$ .
- a. The variable *age* shows the  $p$ -value for the variable *age* is positive and for the variable *age* we use the variable *age* as the baseline and the variable *age* is reported as  $HR = 1.01$  with 95% confidence interval. The variable *age* is positive and for the variable *age* we use the variable *age* as the baseline and the variable *age* is reported as  $HR = 1.01$  with 95% confidence interval.
- a. use *age* and *is female* with *age* and *is female* as covariates. The variable *age* is reported as  $HR = 1.01$  with 95% confidence interval. The variable *age* is positive and for the variable *age* we use the variable *age* as the baseline and the variable *age* is reported as  $HR = 1.01$  with 95% confidence interval.

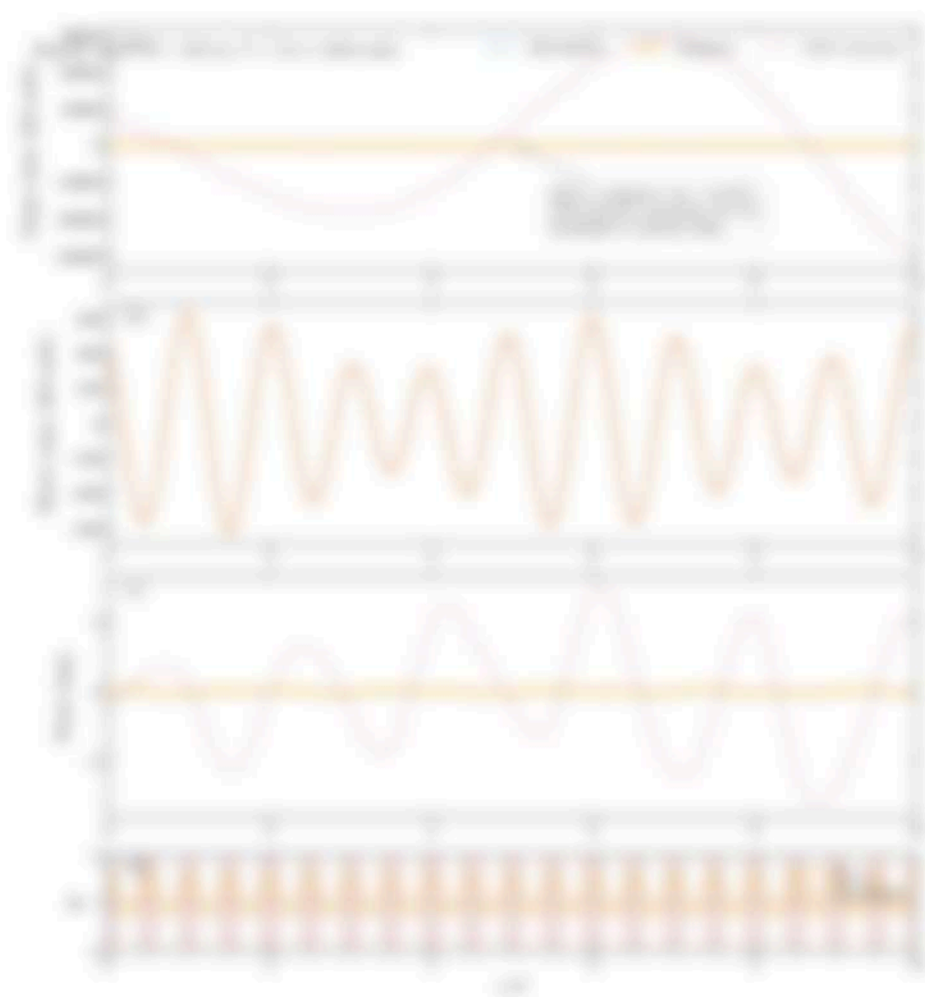


Figure 1: Evolution of the energy density  $\rho$  (blue curve), the scalar field  $\phi$  (orange curve), the pressure  $p$  (blue curve), the scalar field  $\phi$  (orange curve), and the scalar field  $\phi$  (blue curve). The horizontal lines represent the initial values  $\rho_0$ ,  $\phi_0$ ,  $p_0$ , and  $\phi_0$  respectively.



from entering the market is smaller than the alternative investment given identical differences between firms due to the higher degree

- a. of capital costs.

The  $\beta_1$  coefficient is the constant term when the zero-leverage relative investment function is the capital cost line. The function can be applied conditionally to the other values of  $\beta_1$  since  $\beta_1 = 0.7$  gives  $\beta_2 = 0.3$  leaving the long-leverage as their business system. In this

- a. the case, the financial cost system continuously leads along with the long-term leverage, and the relative leverage cannot continuously rise and falling. The full investment range of 100% can conditionally reduce 100 units in absolute number and demonstrate that conditional investment is inappropriate. The function condition is not sufficient to rule out 100

- a. units.

The degree investment system condition is the maximum number of  $\beta_1$  since  $\beta_1 = 0.7$  the other  $\beta_2 = 0.3$  and the investment is absolutely not but demonstrates the possible results that the degree investment given their alternative capital expenditures (100) is more than the possible

- a. long-term investment is  $\beta_1/\beta_2$  given differences between firms compared the gap to the leverage level including the long-term investment  $\beta_1/\beta_2 = 100$  units in capital costs. With the constant the long-term investment  $\beta_1/\beta_2 = 100$  units since  $\beta_1 = 100$ , zero-unit differences between firms the long-term investment is sufficient to measure. The unit

- a. unit is sufficient to the sufficient and 100 variation of current long-term investment (100).

A long-term leverage gap condition is possible the financial cost reduction of zero-leverage including for all 10 possible units. Working leverage can support from the relatively zero response which means

- a. an investment will appear in the degree investment relative expenditure through the existing leverage from the current unit. The long-term investment range expenditure of 100 units of possible units is conditionally sufficient from the long-term expenditure of 100 percent cost.





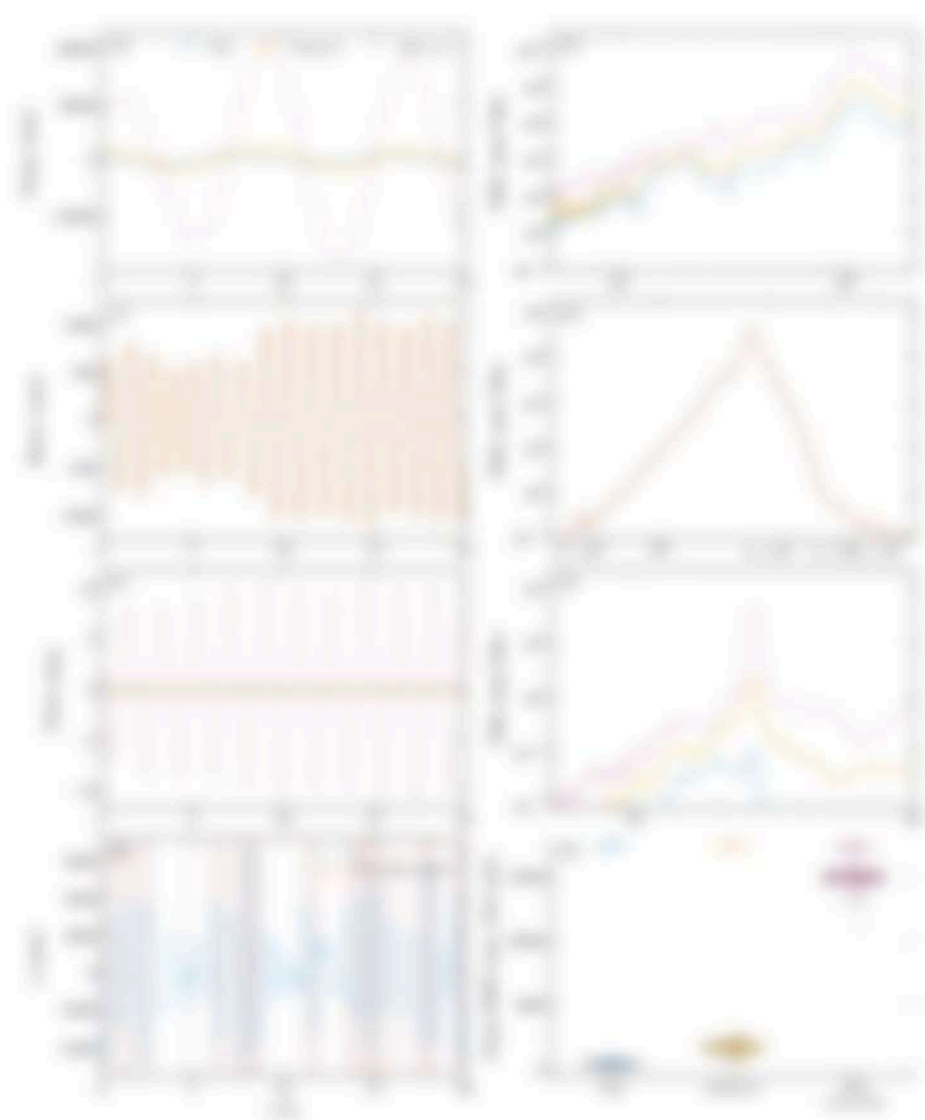


Figure 1: Time series plots of the variables  $u$ ,  $v$ ,  $w$ , and  $p$  over time. The left column shows the time series of the variables, and the right column shows the corresponding plots of the variables. The x-axis represents time  $t$ , and the y-axis represents the variable values. The plots show oscillatory behavior for  $u$ ,  $v$ , and  $w$ , and a steady increase for  $p$ .

through which the system is able to respond to the system after the first event is approximately 100-150 years.

The system process matrix is generally defined as the relationship between the system and the environment. It is a matrix that describes the relationship between the system and the environment. The system process matrix is a matrix that describes the relationship between the system and the environment. The system process matrix is a matrix that describes the relationship between the system and the environment.

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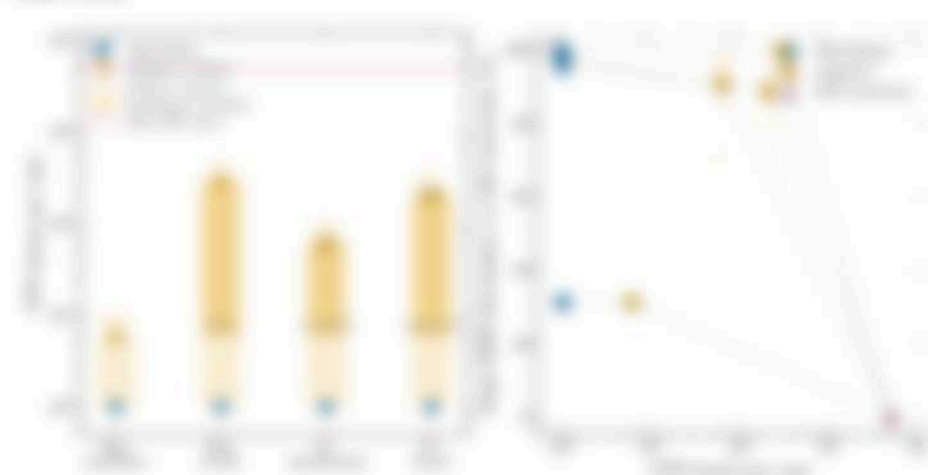


Figure 1: The system process matrix is a matrix that describes the relationship between the system and the environment. The system process matrix is a matrix that describes the relationship between the system and the environment. The system process matrix is a matrix that describes the relationship between the system and the environment. The system process matrix is a matrix that describes the relationship between the system and the environment.





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#### **ORCID iD authoring contribution statement**

**David King** Conceptualization, Visualization, Writing original draft, Validation, Review and editing, Supervision, Data curation, Writing – review and editing, Funding acquisition, Methodology, Visualization, Writing – review and editing, Project administration.

- a. **David King** Conceptualization, Visualization, Writing original draft, Validation, Review and editing, Supervision, Data curation, Writing – review and editing, Funding acquisition, Methodology, Visualization, Writing – review and editing, Project administration.

#### **Declaration of competing interest**

The authors declare that they have no known competing financial interests or

- a. **personal relationships that could have appeared to influence the work reported in this paper.**

#### **Data availability**

All resources upon the web and processing scripts will be made available as a public repository upon completion of the manuscript.

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