

***RAILWAY BRIDGES DAMPING IDENTIFICATION
USING TRAFFIC INDUCED VIBRATION***

- Introduction
- Description of parametric models
- Models validation
- Application to real life signals
- Damping identification along decay
- Conclusion

- Work commissioned and supported by ERRI committee D 214 'Railways bridges for speeds >200 km/h'
- Common methods of damping measurement are :
 - logarithmic decrement or general free decay analysis
 - frequency response under controlled excitation (sinus or transient)
- Because of their low cost, free decay analysis methods are preferred
 - due to the non stationarity of train excitation, signal processing problems occur : spectral methods are inefficient
 - the time methods are well adapted to solve these problems

- Time methods are called parametric models
 - commonly used in speech processing and automatics since the late 60 's
 - the time signal is used to identify model parameters
 - the parametric models proposed here are
 - AR model (Auto Regressive)
 - Eigen value model (Prony Pisarenko)
 - the logarithmic decrement method was also tested

AR model

- sampled general linear system of input e_t and output x_t may be modeled by difference equation

$$x_t + a_1 x_{t-1} + \dots + a_{2m} x_{t-2m} = e_t \qquad x_t = - \sum_{k=1}^{2m} a_k x_{t-k} + e_t$$

or in matrix form

$$\begin{bmatrix} x_{2m+1} \\ x_{2m+2} \\ \vdots \\ x_t \end{bmatrix} = \begin{bmatrix} x_{2m} & x_{2m-1} & \cdots & x_1 \\ x_{2m+1} & x_{2m} & \cdots & x_2 \\ \vdots & \vdots & & \vdots \\ x_{t-1} & x_{t-2} & \cdots & x_{t-2m} \end{bmatrix} \begin{bmatrix} -a_1 \\ -a_2 \\ \vdots \\ -a_{2m} \end{bmatrix} + \begin{bmatrix} e_{2m+1} \\ e_{2m} \\ \vdots \\ e_t \end{bmatrix}$$

- a_k are called the AR coefficients
- These coefficients are related to the frequencies ω_k and dampings ζ_k by the relationships (Δt =sampling interval)

$$\omega_k \zeta_k = -\frac{\text{Ln}(|a_k|^2)}{2\Delta t}$$

$$\omega_k \sqrt{1 - \zeta_k^2} = \frac{\Delta t \tan(\Im(a_k) / \Re(a_k))}{\Delta t}$$

- The solution is found through least square estimation

$$X = \Phi_x \theta + E \qquad \theta = \left(\Phi_x^T \Phi_x \right)^{-1} \Phi_x^T X$$

- One cannot fully justify the use of the AR model unless the input of the system consists in white noise, which is not the case here
- The least square estimation is a source of bias errors in parameter estimation

The eigen value method (Prony-Pisarenko)

- When no input is present system equation can be written

$$\text{as } x_t + \sum_{k=1}^{2m} a_k x_{t-k} = 0$$

$$[X \ \Phi_x] \begin{bmatrix} 1 \\ -\theta \end{bmatrix} = \tilde{\Phi}_x \begin{bmatrix} 1 \\ -\theta \end{bmatrix} = \{0\}$$

$$\tilde{\Phi}_x^T \tilde{\Phi}_x \begin{bmatrix} 1 \\ -\theta \end{bmatrix} = R_{xx} \begin{bmatrix} 1 \\ -\theta \end{bmatrix} = \{0\}$$

- R_{xx} is the covariance matrix
 - parameter vector $[1 - \theta]^T$ is the covariance matrix eigen vector associated with eigen value 0
 - as R_{xx} is definite semi positive, this eigen value is the lowest one
 - it can be shown that in presence of noise at the output, the eigen value remains the lowest one

- The method simply consists in computing the covariance matrix R_{xx} and then calculating eigen vectors. The eigen vector whose eigen value is the lowest gives the a_k parameters

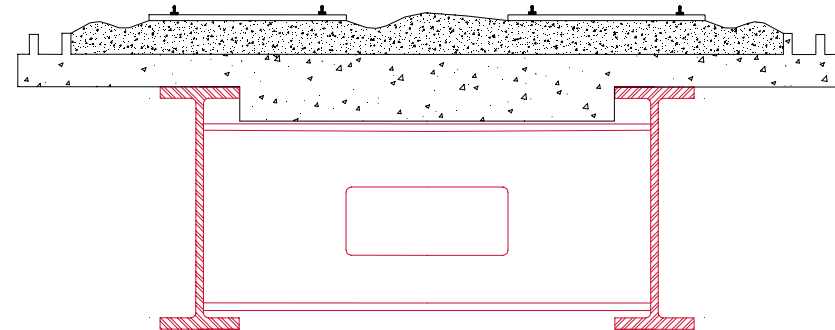
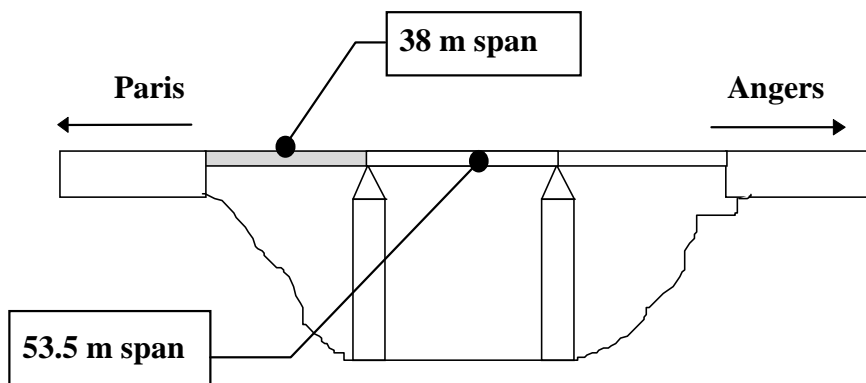
- Simulations have been performed to validate the different models and estimate errors in damping estimation
 - AR, eigen model and logarithmic decrement have been tested
 - all methods are sensitive to
 - size (number of samples) of data processed
 - sampling frequency
 - filtering
 - modal coupling
 - The simulation aim is to optimize the choice of signal processing parameters

- logarithmic decrement
 - bias errors are small $< 5\%$
 - more than 10 periods of signal are necessary to obtain less than 15 % random error
 - in the case of several modes, the filtering around each mode gives similar results with less than 20 % random error
 - results are coherent when the modes are sufficiently separated
- AR & Eigen models
 - ~10 periods of signal are sufficient
 - AR model is biased but the filtering reduces this effect
 - eigen model is unbiased ($< 5\%$)
 - in these conditions bias and random errors are less than 10 %
 - with three modes bias error may increase (15 %) when coupling between modes is important (4% damping)

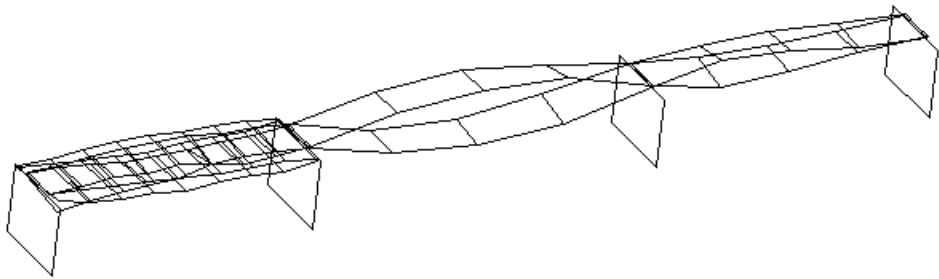
■ Conclusion

- logarithmic decrement is a good estimator, if the modes are not too close one to the other
- AR & Eigen models give good results, The Eigen model gives better results than AR model since it is unbiased
- simulations gave confidence in time methods. A strategy of mode identification is well defined and may be applied to real life measurements
- as errors in simulations are less than 15 %, greater errors must be expected in reality

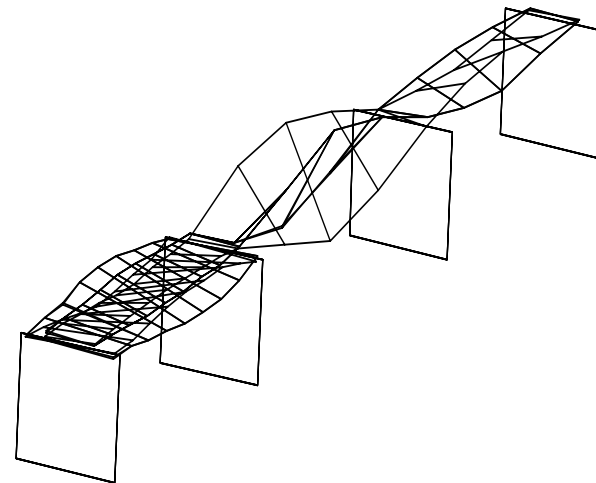
- Train passing by were recorded on Briollay bridge on TGV Atlantique high speed line
 - 3 spans steel/concrete bridge



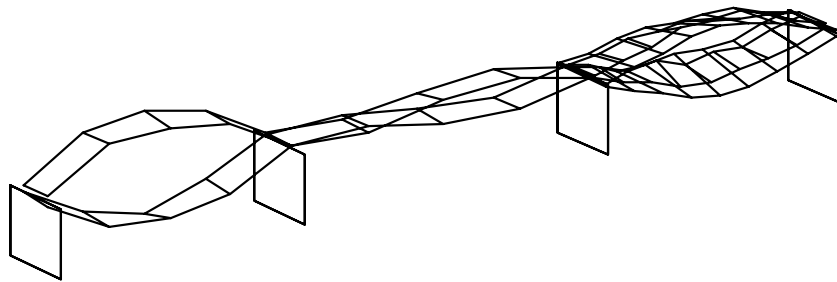
- Modal analysis was performed with hammer excitation



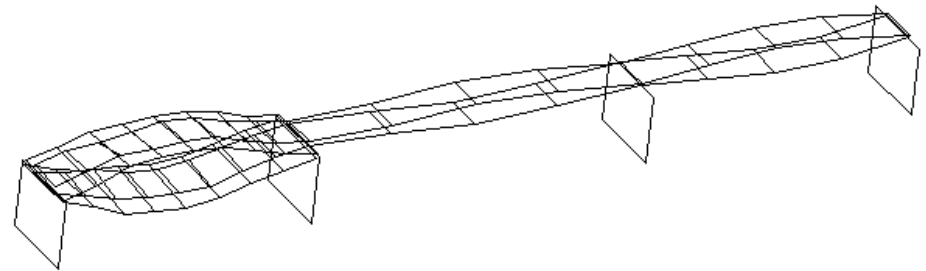
whole bridge flexion mode @ 2.29 Hz



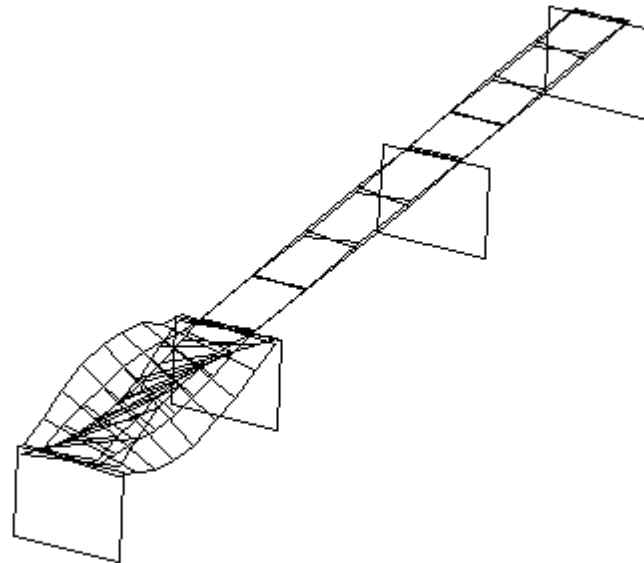
whole bridge torsion mode @ 3.28 Hz



whole bridge flexion mode @ 3.66 Hz



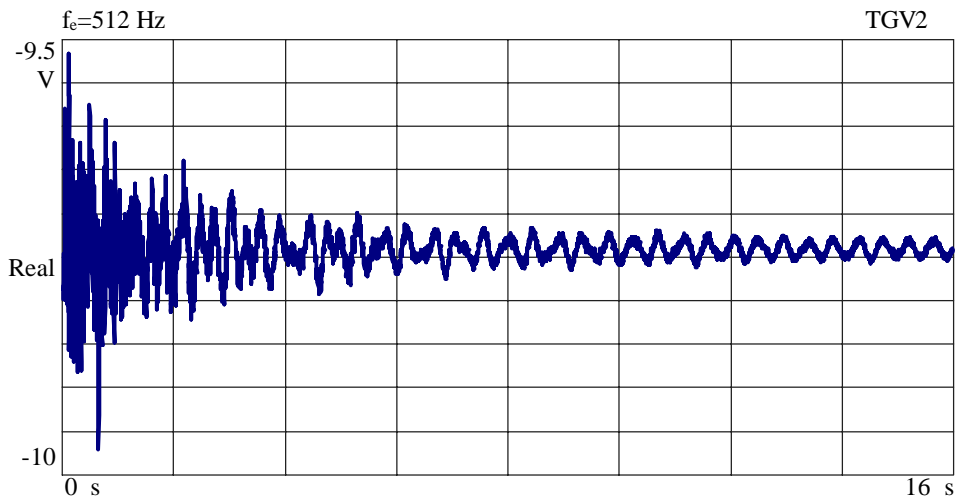
1stspan flexion mode @ 4.16 Hz



1stspan torsion mode @ 4.96 Hz

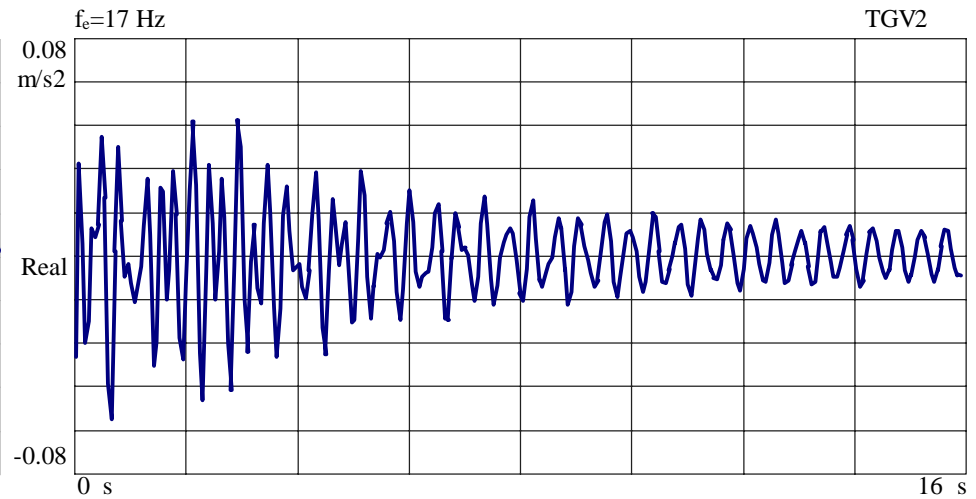
- 6 train passing by are analyzed
 - 5 TGV
 - 1 TER
- Strategy for analyzing decay signals
 - measure the spectral components of the signal
 - separate modes to be identified in small groups
 - for each group, apply optimum sampling and filtering to isolate modes
 - identify frequency and damping

EURODYN '99 *Application to real life signals*



measurement signal

$f_e = 512$ Hz



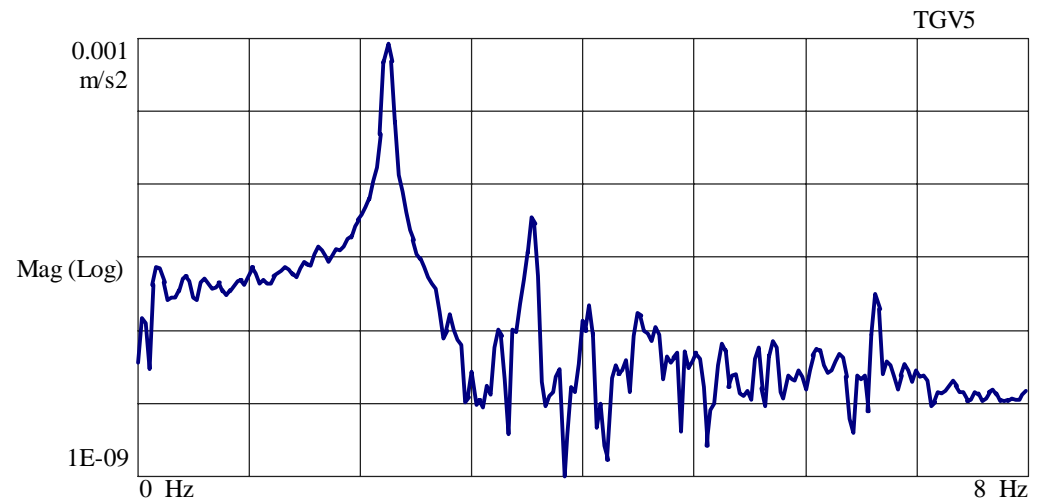
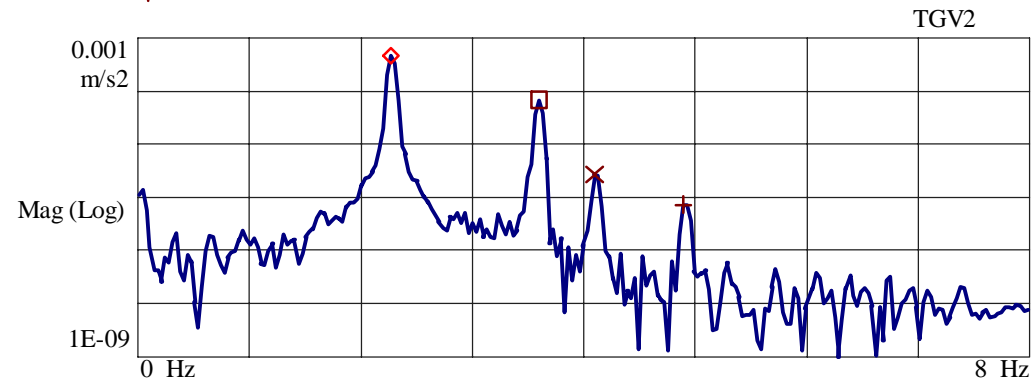
30 times decimated signal

$f_e = 17$ Hz

■ Spectra

- very noisy
- 2 leading modes at 2.3 & 3.6 Hz
- 2 other modes at 4.1 & 4.9 Hz (not always excited)
- first torsion mode at 3.3 Hz not excited

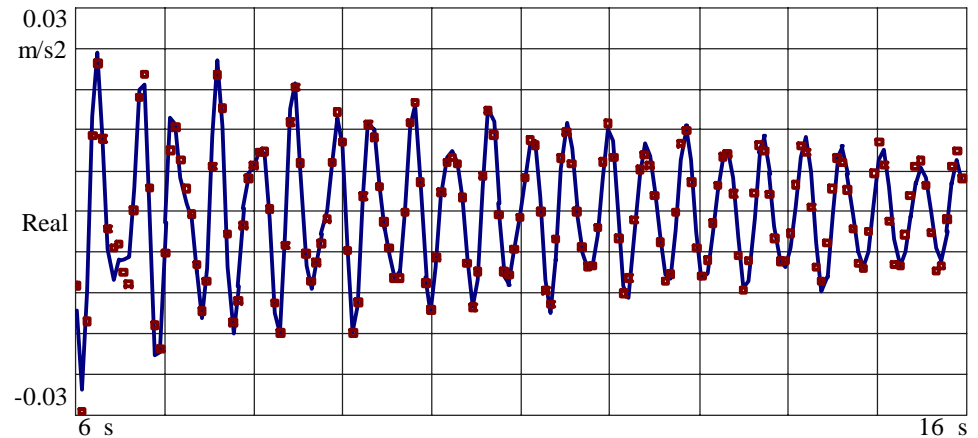
◇ X:2.2644 Hz Y:469.4962 $\mu\text{m/s}^2$
 □ X:3.5964 Hz Y:68.70036 $\mu\text{m/s}^2$
 × X:4.0959 Hz Y:2.729211 $\mu\text{m/s}^2$
 + X:4.8951 Hz Y:735.5877 nm/s^2



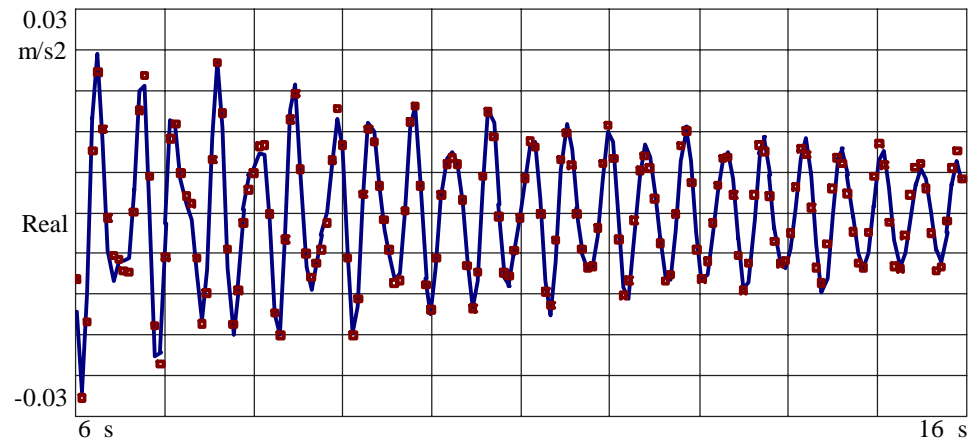
■ identification with AR and Eigen model

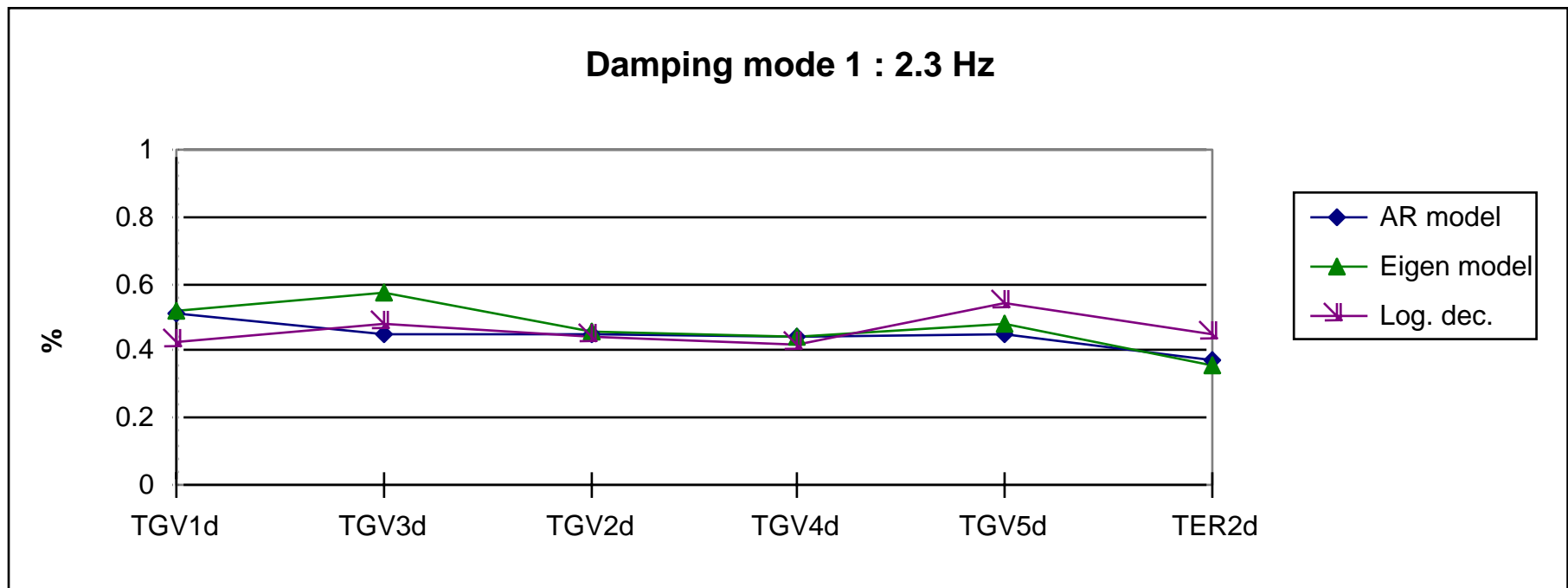
- red points give the synthetized curve with model results
- a fit error is computed and gives model confidence

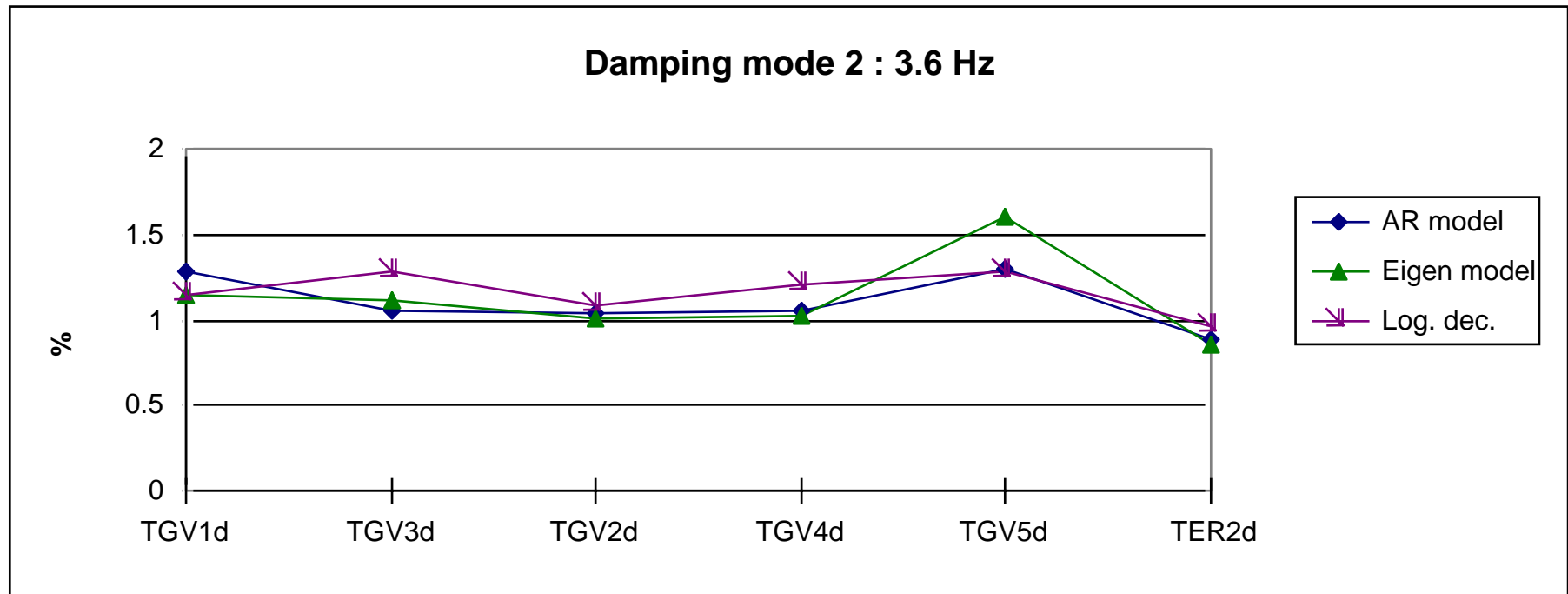
AR model identification

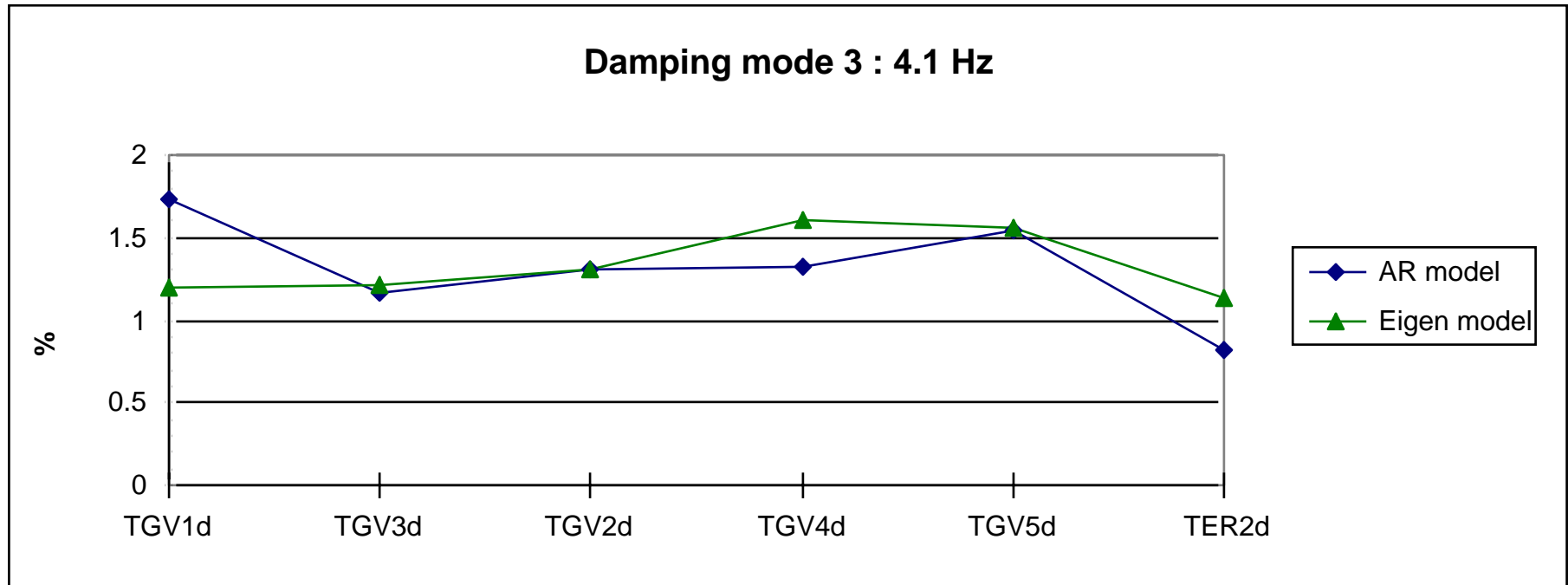


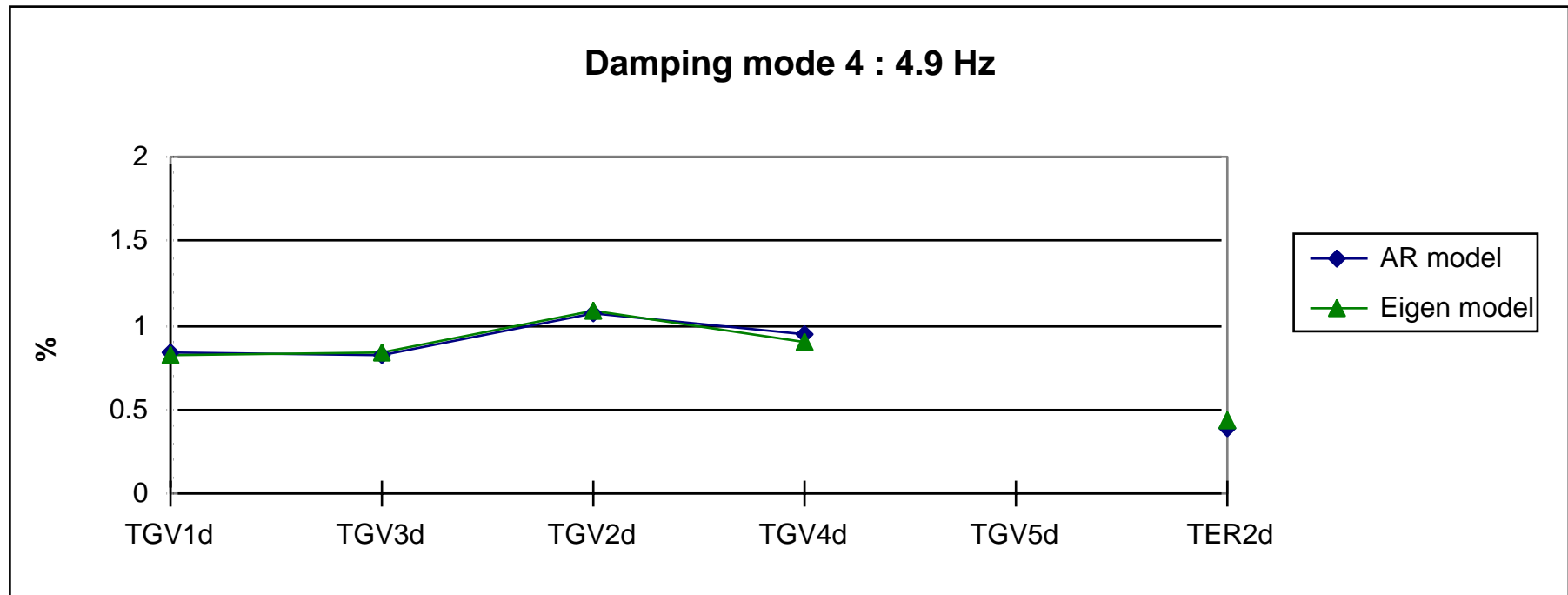
Eigen model identification









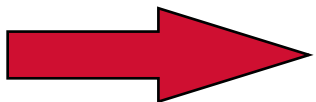


	AR model		Eigen model		logarithmic decrement	
	damp. (%)	scatter (%)	damp. (%)	scatter (%)	damp. (%)	scatter (%)
mode 1 2.3 Hz	0.45	10	0.47	15	0.46	10
mode 2 3.6 Hz	1.1	15	1.13	22	1.16	11
mode 3 4.1 Hz	1.32	24	1.33	15	not	
mode 4 4.9 Hz	0.81	32	0.81	30	computable	

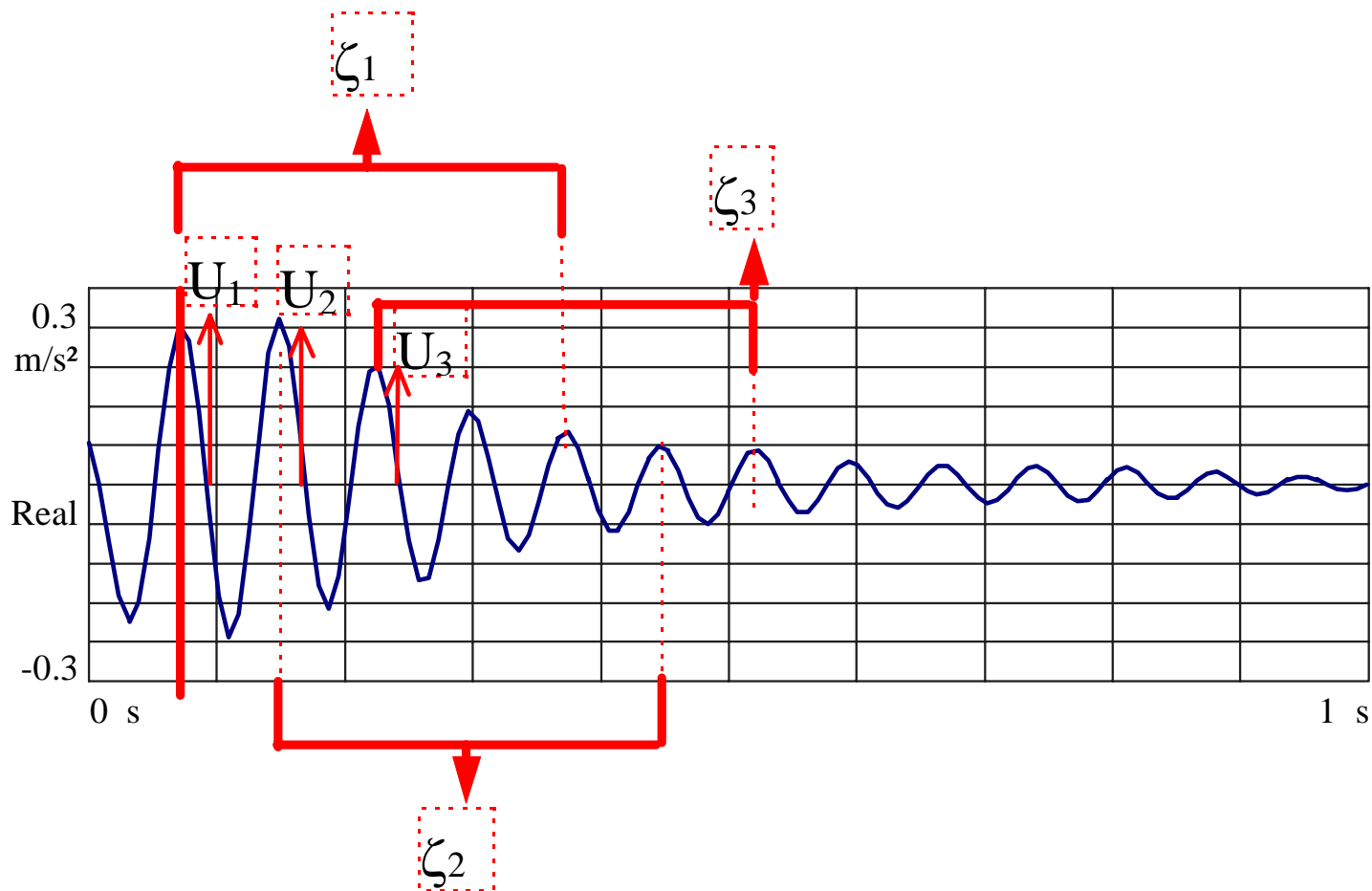
■ Result synthesis

- frequencies are estimated with a very good accuracy ~1 %
- modes 1 & 2 are estimated with maximum 10-20 % dispersion
- modes 3 & 4 are estimated with maximum 30 % dispersion
- discrepancies are caused by measurement/estimation errors AND bridge and excitation characteristics (non linearities, train loading)
- as modes 3 & 4 are only slightly emerging from noise, dispersion is greater

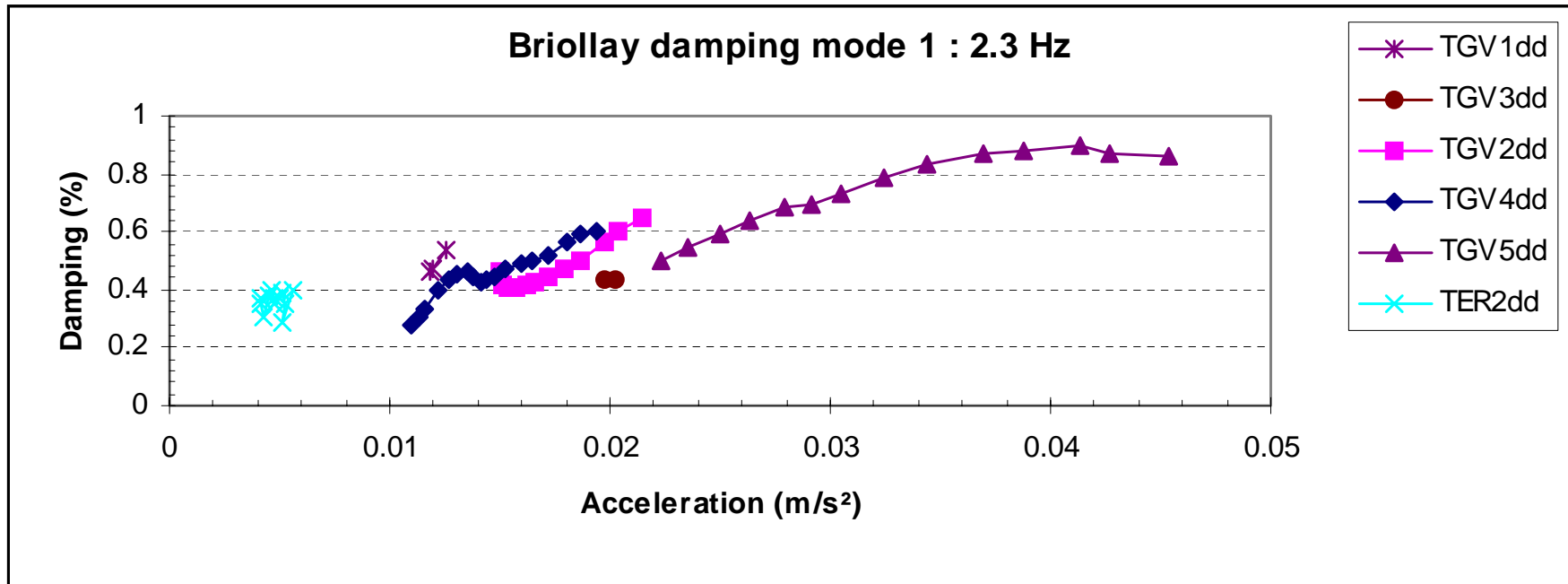
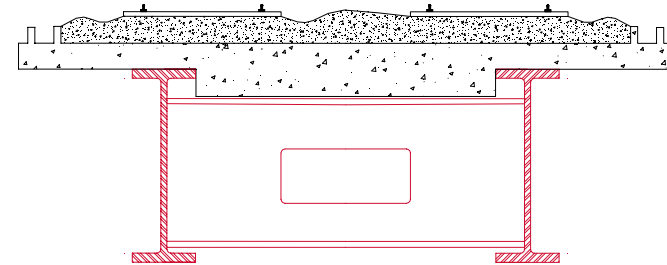
- Due to non linearities, **damping varies along decay** : the change in damping can be estimated using the following procedure
 - the damping is estimated using the Prony method in steps of ~ 10 periods (filtering 48 dB/oct. at 1.5 X first frequency when possible)
 - maximum of amplitude is measured : this amplitude is associated with a damping value



Plot of the damping as a function of amplitude

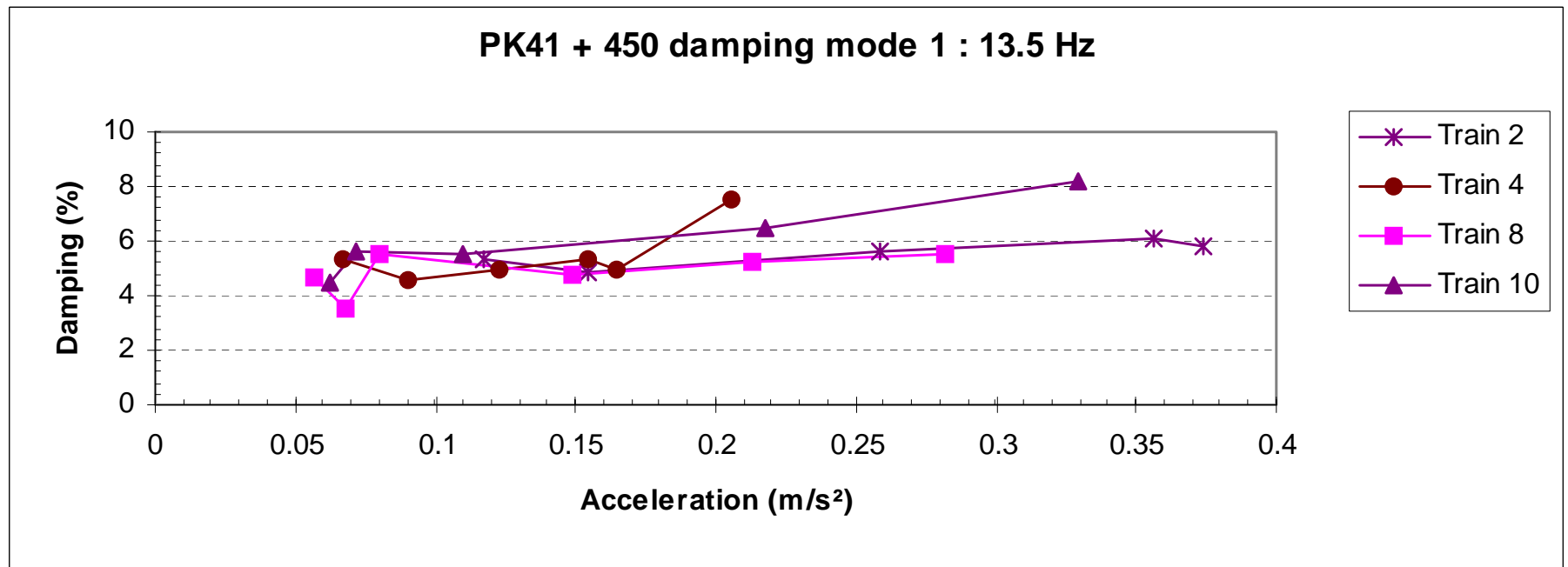
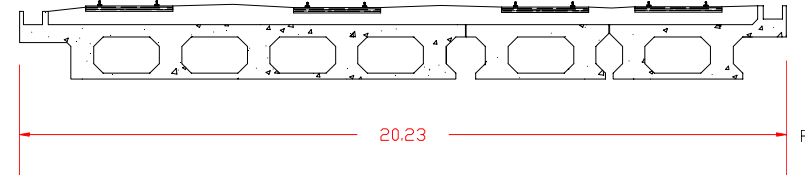


- 3 spans steel/concrete decks
 - strong dependance on amplitude
 - $0.3 < \zeta < 0.9 \%$

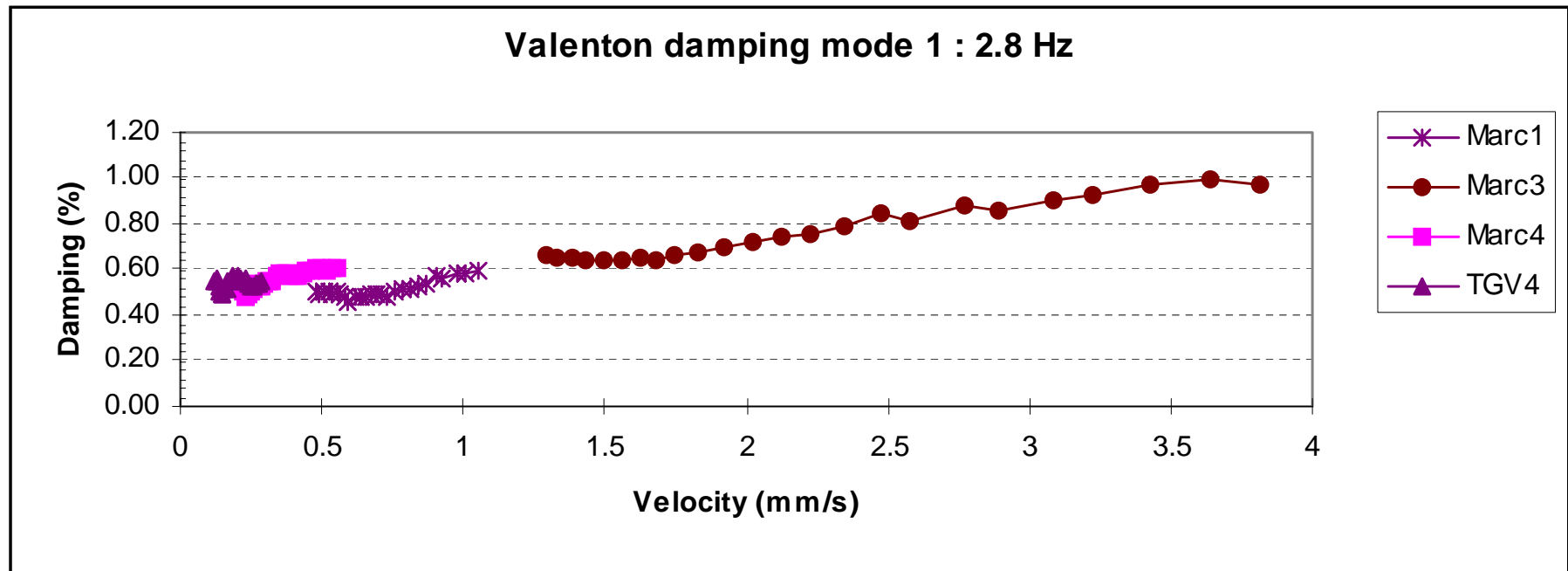
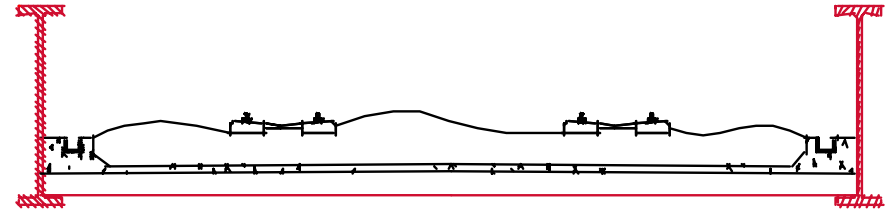


■ 16.5 m long reinforced concrete deck

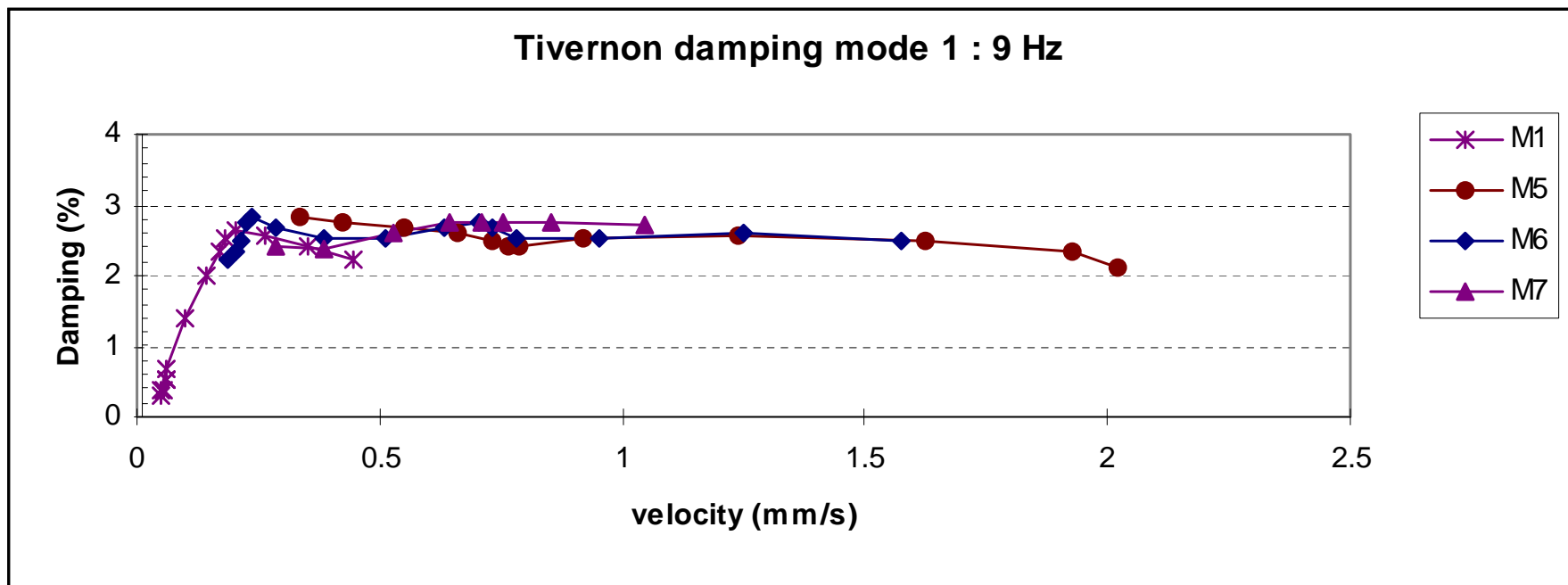
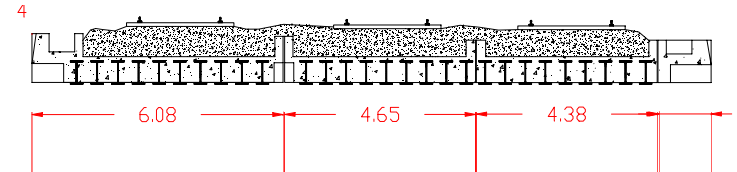
- $4 < \zeta < 8 \%$
- slight increase with acceleration



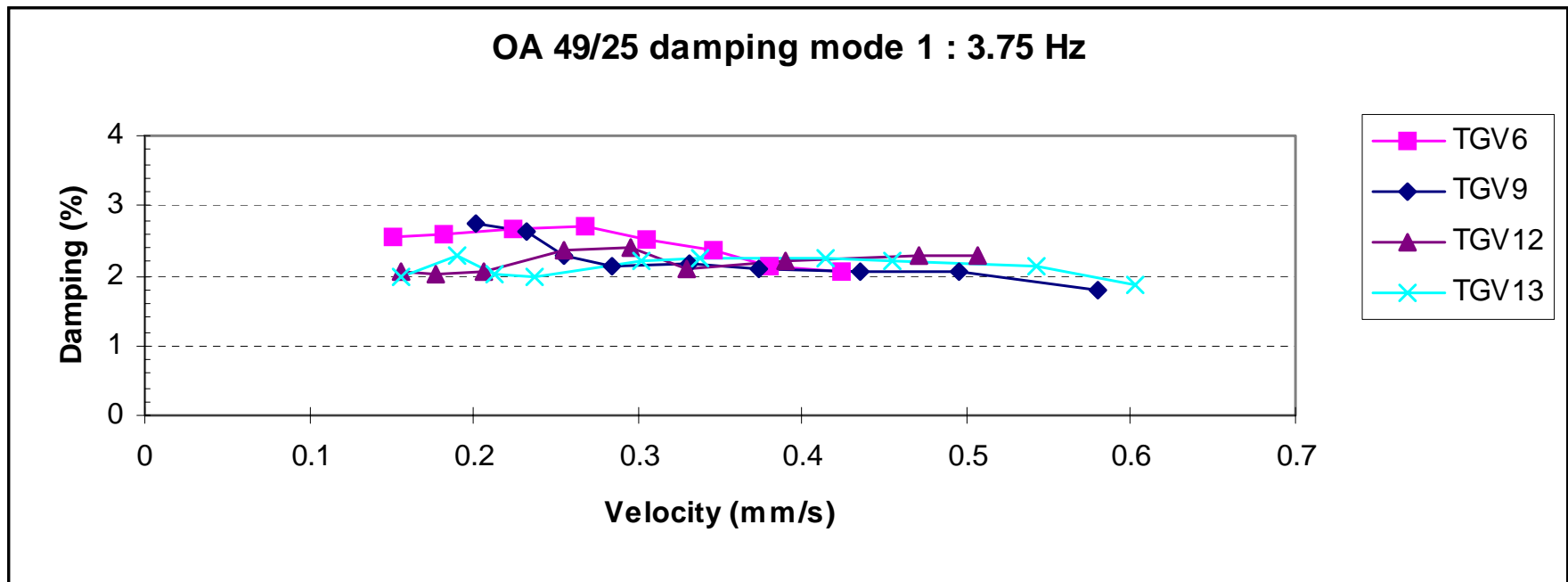
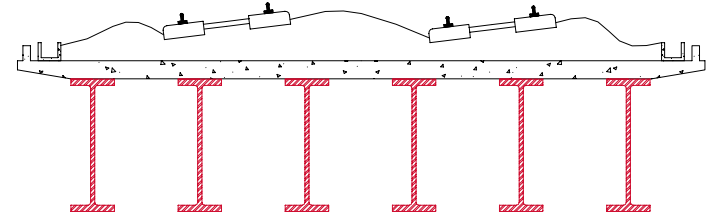
- 46 m long steel deck
 - strong dependance on amplitude
 - good correlation between trains



- 3 spans 11.7 m long steel/concrete decks
 - constant damping (except for 1 train)
 - good correlation between trains

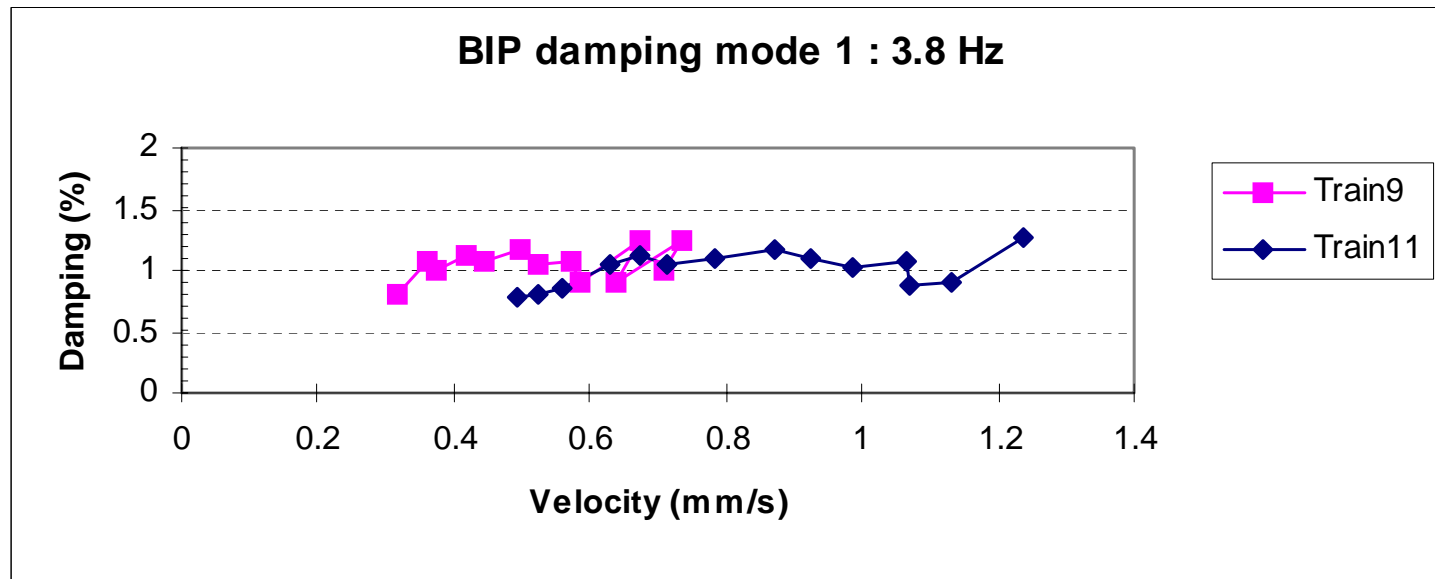
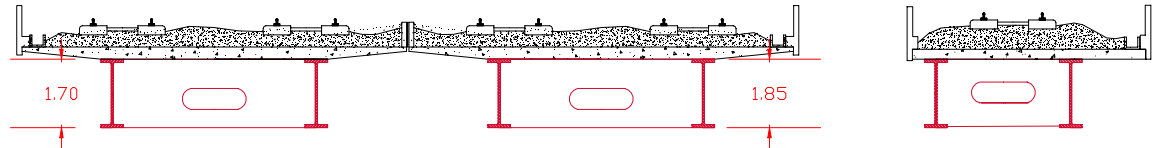


- 46 m long steel/concrete deck
 - constant damping
 - good correlation between trains



- 2 spans 33-35 m long steel/concrete decks

- damping constant



■ Conclusion

- two types of behaviour
 - non linearity : damping increases linearly with amplitude
 - constant damping
- due to the deformation amplitude, different train characteristics and speeds lead to different results
 - Briollay : TER 0.25 % ; TGV 0.4 -> 1%
- This kind of analysis gives good confidence on measurements and damping values
 - scattering in measurements can be explained

- Prony and AR models have proved to be very efficient
 - good accuracy in results
 - coupled modes identification
- Recommendations in damping estimation
 - proper signal filtering is of great help
 - **damping has to be estimated by a statistical approach**
 - use several methods (log. dec. AR eigen method) in order to allow assessment of errors made in the estimation procedure
 - use several measurements to estimate random errors
 - mean and standard deviation must be given

- as damping may vary with vibration level
 - it should be studied throughout the decay period (when possible)
 - damping values must be associated with a deck vibration amplitude
 - trains type and speeds are of great importance in assessing the reproductibility of results