

# Operational modal analysis for Canton Tower

Yan Niu<sup>1,2</sup>, Peter Kraemer<sup>3</sup> and Claus-Peter Fritzen<sup>\*1,2</sup>

<sup>1</sup>Center for Sensor Systems (ZESS), University of Siegen, Germany

<sup>2</sup>Institute of Mechanics and Control Engineering-Mechatronics, University of Siegen, Germany

<sup>3</sup>Woelfel Beratende Ingenieure, Hoechberg, Germany

(Received February 17, 2012, Revised March 21, 2012, Accepted April 25, 2012)

**Abstract.** The 610 m high Canton Tower (formerly named Guangzhou New Television Tower) is currently considered as a benchmark problem for structural health monitoring (SHM) of high-rise slender structures. In the benchmark study task I, a set of 24-hour ambient vibration measurement data has been available for the output-only system identification study. In this paper, the vector autoregressive models (ARV) method is adopted in the operational modal analysis (OMA) for this TV tower. The identified natural frequencies, damping ratios and mode shapes are presented and compared with the available results from some other research groups which used different methods, e.g., the data-driven stochastic subspace identification (SSI-DATA) method, the enhanced frequency domain decomposition (EFDD) algorithm, and an improved modal identification method based on NExt-ERA technique. Furthermore, the environmental effects on the estimated modal parameters are also discussed.

**Keywords:** operational modal analysis (OMA); Canton Tower, vector autoregressive models (ARV) method; environmental effects; structural health monitoring (SHM)

---

## 1. Introduction

According to (Farrar and Worden 2007), structural health monitoring (SHM) is defined as the process of implementing a damage identification strategy for aerospace, civil and mechanical engineering infrastructures. It is clear that the modal properties (i.e., natural frequencies, damping ratios and mode shapes) play an important role in the SHM study. For example, in vibration-based damage assessment, the modal properties from both the undamaged and damaged structures can be used to update the structural model in order to locate and quantify the damage. Besides this, with the help of the modal properties, an updated and reliable structural model can also benefit the model-based load reconstruction study, e.g., the method in (Niu *et al.* 2011a). In structural control design, an appropriate tuning frequency of tuned mass damper (TMD) has a close relation with the structural modal frequencies (Ni *et al.* 2009). The above mentioned three points, to some extent, can reflect the importance of modal properties. It is known that the operational modal analysis (OMA), or named as output-only modal analysis, can identify the modal characteristics using only the response measurements of structures in their operational conditions subject to ambient or natural excitation (Zhang *et al.* 2005).

---

\*Corresponding author, Professor, E-mail: [fritzen@imr.mb.uni-siegen.de](mailto:fritzen@imr.mb.uni-siegen.de)

The Canton Tower, formerly called Guangzhou New Television Tower (GNTVT), has a total height of 610 m, including a 454 m high main tower and a 156 m high antenna mast, and is currently considered as the landmark of the Guangzhou city in southern China. It serves for the functions of TV and radio transmission, sightseeing, cultural entertainment, and offers an orbital Ferris wheel, a ceremony hall, observatory decks, 4D cinemas, revolving restaurants, skywalk, etc. In November 2010, the Canton Tower also took part in broadcasting the Asian Games hosted by Guangzhou. To maintain the safe and reliable operation of the Canton tower, researchers from the Hong Kong Polytechnic University extended their practice and experiences gained in developing structural health monitoring (SHM) systems for bridges to high-rise structures, and have designed and implemented a long-term SHM system for the Canton Tower (Ni *et al.* 2011). Meanwhile, to ensure and improve the comfort of people staying in this tower, a hybrid mass damper (HMD) control system is installed on the main tower and two tuned mass dampers (TMDs) are suspended on the antenna mast to mitigate the wind-induced vibration (Ni and Zhou 2010).

Under the auspices of Asian-Pacific Network of Centers for Research in Smart Structures Technology (ANCRiSST), an SHM benchmark problem for high-rise structures is being developed by taking the instrumented Canton Tower as the host structure (Ni *et al.* 2012, Xia *et al.* 2009). The benchmark study task I is output-only system identification and FE model updating. At present, a set of 24 hour field measurement data and a reduced-order FE model for the Canton Tower have been available (Lin *et al.* 2010). This paper focuses on the output-only system identification part using the available field measurement data.

In 2010, Kraemer and Fritzen applied vector autoregressive models (ARV) method for the study of dynamic characteristics of offshore wind energy plants (Kraemer and Fritzen 2010a, Kraemer and Fritzen 2010b). In this contribution, the ARV technique is also adopted in the operational modal analysis for the Canton Tower. The organization of this paper is as follows. In Section 2, a short description of the field measurement data is provided. In Section 3, the principles of the ARV method are briefly reviewed. In Section 4, the identified natural frequencies, damping ratios and mode shapes are presented and compared with the available results presented by different research groups (Chen *et al.* 2011a, Chen *et al.* 2011b, Faravelli *et al.* 2010a, Faravelli *et al.* 2010b, Loh *et al.* 2011, Niu *et al.* 2011b, Niu *et al.* 2011c, Ye *et al.* 2011). In Section 5, the environmental effects on the identified modal parameters are discussed. Finally, the conclusions are given.

## 2. Field measurement data for OMA

The 454 m high main tower of the Canton Tower has a tube-in-tube geometry, consisting of a reinforced concrete inner structure and a steel lattice outer structure, as shown in Fig. 1. The 156 m high antenna mast mounted on the top of the main tower is a steel spatial structure. Detailed dimension and configuration description can be found in (Ni *et al.* 2009). In Task I of this benchmark problem, twenty uni-axial accelerometers (Tokyo Sokusin AS-2000C), an anemometer (RM Young, 05103L) and a thermocouple (PT1000) were employed from the designed SHM system. The positions and measurement directions of these sensors are illustrated in Figs. 2 and 3.

The acceleration, wind direction and wind speed were measured with sampling frequency 50 Hz, and the temperature was sampled every minute (1/60 Hz). The field measurement data available for the OMA were recorded permanently during 24 hours from 18:00 h on January 19, 2010, to 18:00 h

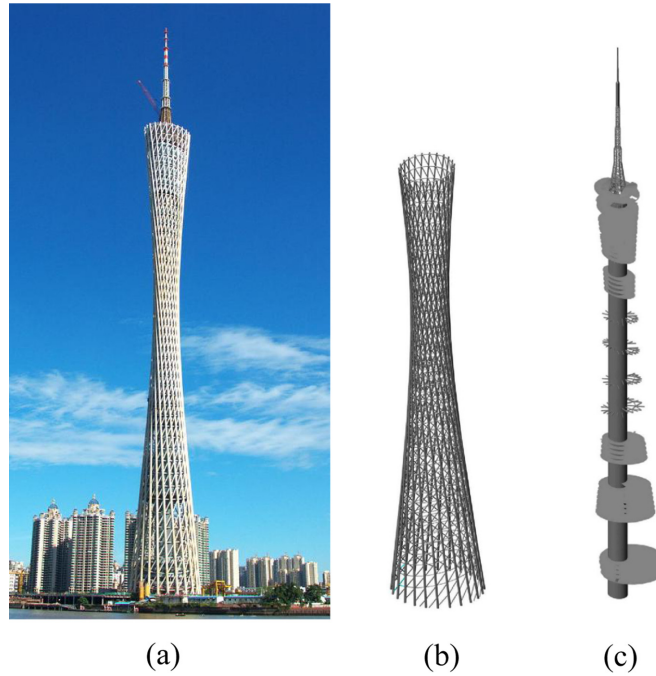


Fig. 1 (a) Canton tower, (b) outer steel structure, and (c) inner concrete structure (Source: The Hong Kong Polytechnic University)

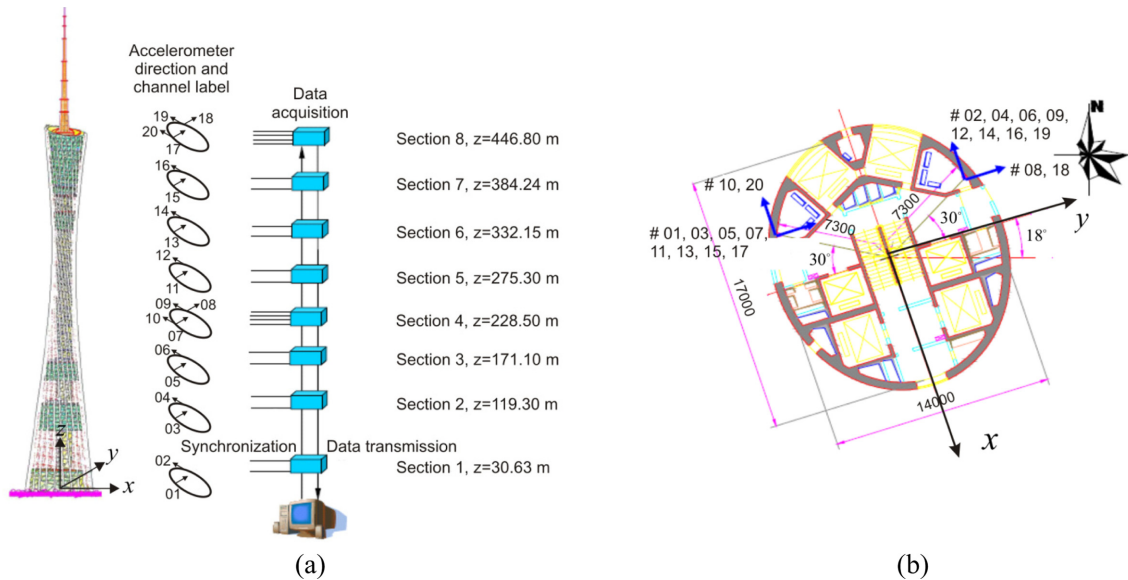


Fig. 2 (a) Positions of accelerometers and data acquisition system and (b) measurement directions of acceleration and channel labels (Source: The Hong Kong Polytechnic University)

on January 20, 2010. More detailed information on the field measurements and sensor specifications are presented in (Lin *et al.* 2010).

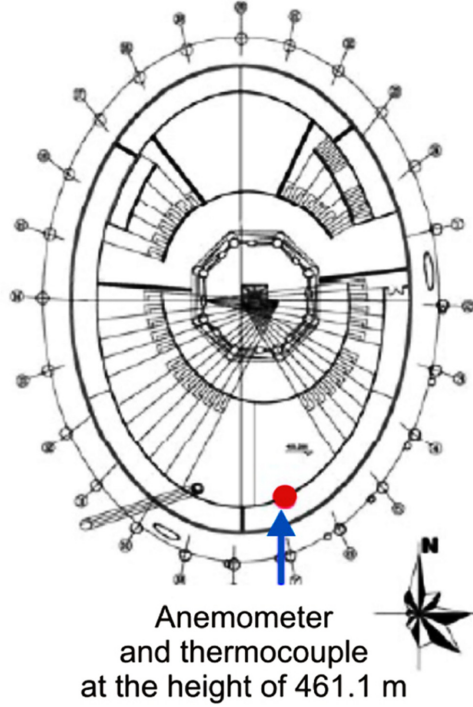


Fig. 3 Positions of the anemometer and the thermocouple (Source: The Hong Kong Polytechnic University)

### 3. Vector autoregressive models (ARV) method

The measurements obtained at equally spaced time instants  $k$  can be modeled by an arbitrary order vector autoregressive model (ARV) of the following form

$$\mathbf{y}_k = \mathbf{b} + \sum_{l=1}^p \mathbf{A}_l \mathbf{y}_{k-l} + \boldsymbol{\varepsilon}_k \quad (1)$$

where  $\mathbf{y}_k$  denotes the measurement vector from  $m$  sensors at time instant  $k$ ,  $p$  is the order of this AR model and  $\mathbf{A}_1, \dots, \mathbf{A}_p \in \mathbb{R}^{m \times m}$  are coefficient matrices.  $\mathbf{b}$  is an intercept vector.  $\boldsymbol{\varepsilon}_k$  is assumed to be an uncorrelated zero mean random vector, representing the residual between real measurements and the model at time instant  $k$ . Assuming  $\mathbf{b}$  is a zero vector, the  $p$ -th order AR model (AR( $p$ )) can be transformed to a state space form as in Eqs. (2(a) and (b)) with expressions in Eqs. (3)-(6).

$$\mathbf{x}_{k+1} = \mathbf{A}_d \mathbf{x}_k + \mathbf{w}_k \quad (2a)$$

$$\mathbf{y}_k = \mathbf{C} \mathbf{x}_k \quad (2b)$$

$$\mathbf{x}_k = \begin{bmatrix} \mathbf{y}_k^T & \mathbf{y}_{k-1}^T & \dots & \mathbf{y}_{k-p+1}^T \end{bmatrix}^T \in \mathbb{R}^{(mp) \times 1} \quad (3)$$

$$\mathbf{A}_d = \begin{bmatrix} \mathbf{A}_1 & \mathbf{A}_2 & \dots & \mathbf{A}_{p-1} & \mathbf{A}_p \\ \mathbf{I} & \mathbf{0} & \dots & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{I} & \dots & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \ddots & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \dots & \mathbf{I} & \mathbf{0} \end{bmatrix} \in R^{(mp) \times (mp)} \quad (4)$$

$$\mathbf{w}_k = \begin{bmatrix} \boldsymbol{\varepsilon}_{k+1}^T & \mathbf{0} & \dots & \mathbf{0} \end{bmatrix}^T \in R^{(mp) \times 1} \quad (5)$$

$$\text{and } \mathbf{C} = \begin{bmatrix} \mathbf{I} & \mathbf{0} & \dots & \mathbf{0} \end{bmatrix} \in R^{m \times (mp)} \quad (6)$$

The robust calculation of the autoregressive coefficient matrices takes advantage of the ARFIT MATLAB package (Neumaier and Schneider 2001, Schneider and Neumaier 2001). Then the dynamical characteristics of the investigated structure can be extracted by solving the eigenvalue problem of the discrete state space system matrix  $\mathbf{A}_d$

$$\mathbf{A}_d = \boldsymbol{\Psi} \boldsymbol{\Lambda} \boldsymbol{\Psi}^{-1} \quad (7)$$

where  $\boldsymbol{\Lambda}$  is a diagonal matrix of associated discrete-time eigenvalues  $\lambda_{p,i}^d$  ( $i=1, \dots, mp$ ) and  $\boldsymbol{\Psi}$  contains the corresponding eigenvectors  $\boldsymbol{\Psi}_{p,i}$  ( $i=1, \dots, mp$ ) in its  $i$ -th column. Using Eq. (8),  $\lambda_{p,i}^d$  can be transformed to its continuous-time counterpart

$$\lambda_{p,i}^c = (\ln \lambda_{p,i}^d) / \Delta t \quad (8)$$

where  $\Delta t$  is the sampling time. The corresponding natural frequency  $f_{p,i}$ , damping ratio  $\zeta_{p,i}$  and complex mode shape  $\boldsymbol{\varphi}_{p,i}$  ( $i=1, \dots, mp$ ) can be calculated using Eqs. (9)-(12).

$$\lambda_{p,i}^c, \lambda_{p,i}^{c*} = -\zeta_{p,i} \omega_{p,i} \pm j \omega_{p,i} \sqrt{1 - \zeta_{p,i}^2} \quad (9)$$

where  $\lambda_{p,i}^{c*}$  is the complex conjugate of  $\lambda_{p,i}^c$ .

$$f_{p,i} = |Im(\lambda_{p,i}^c)| / (2\pi) \quad (10)$$

$$\zeta_{p,i} = -Re(\lambda_{p,i}^c) / |\lambda_{p,i}^c| \quad (11)$$

$$\boldsymbol{\varphi}_{p,i} = \begin{bmatrix} \mathbf{0}_{m \times m} & \mathbf{0}_{m \times m} & \dots & \mathbf{I}_m \end{bmatrix}_{m \times (mp)} \boldsymbol{\Psi}_{p,i} \quad (12)$$

In order to separate the physical modes from the spurious ones caused by noise and numerical errors, the percentage frequency difference  $f_{diff}$ , the percentage damping ratio difference  $d_{diff}$  and the modal assurance criterion (MAC) value between the modal properties from the AR( $p$ ) model and those from the AR( $p-1$ ) model are adopted, and the following three criteria are introduced

$$(a) f_{diff} = \frac{|f_{p,i} - f_{p-1,i}|}{f_{p-1,i}} \leq f_{tol} \quad (13)$$

$$(b) \ d_{diff} = \frac{|d_{p,i} - d_{p-1,j}|}{d_{p-1,j}} \leq d_{tol} \quad (14)$$

$$(c) \ MAC = \frac{|\boldsymbol{\varphi}_{p,i}^T \boldsymbol{\varphi}_{p-1,j}^*|^2}{\boldsymbol{\varphi}_{p,i}^T \boldsymbol{\varphi}_{p,i}^* \boldsymbol{\varphi}_{p-1,j}^T \boldsymbol{\varphi}_{p-1,j}^*} \geq MAC_{tol} \quad (15)$$

where  $p \geq 2$ ,  $(i=1, \dots, mp)$ ,  $j=1, \dots, m(p-1)$ ,  $\boldsymbol{\varphi}_{p,i}^T$  is the transpose of  $\boldsymbol{\varphi}_{p,i}$ , and  $\boldsymbol{\varphi}_{p,i}^*$  is the conjugate transpose of  $\boldsymbol{\varphi}_{p,i}$ .

#### 4. OMA results

It is known that the monitoring data segment length should be long enough to reduce the noise effects, while more data, on the other hand, lead to longer computation time. In accordance with the choice in (Faravelli *et al.* 2010a, Faravelli *et al.* 2010b), 1 h is selected as the time interval in this paper for each data segment. However, the originally provided 24 measurements of 1 h duration are not used directly for the present study but first decomposed into 70 overlapping data sets of 1 h duration with a 20 min shift, as shown in Fig. 4.

The reason for such an operation is to improve the continuity of the one hour mean wind speed and the one hour mean temperature in later environmental effect discussion. For a comparison of

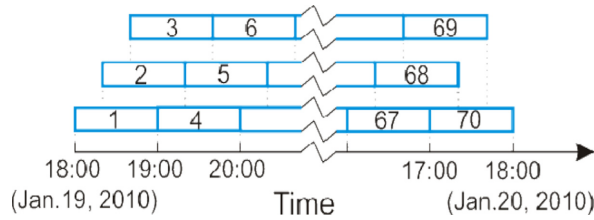


Fig. 4 20 min shift data operation

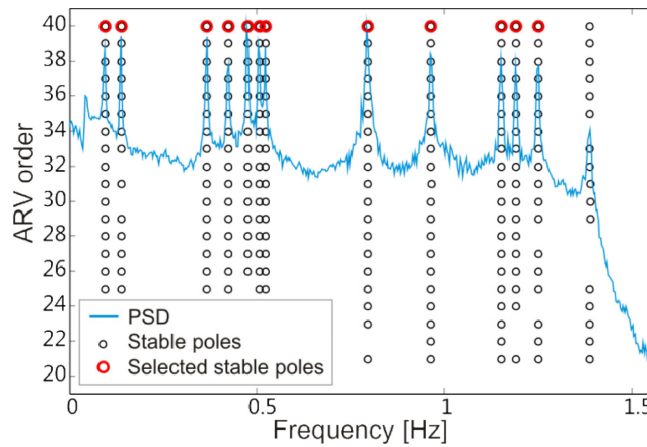


Fig. 5 Stabilization diagram for data set no. 61

Table 1 Identified natural frequencies and damping ratios

Mode Nr.	Natural frequency $f$ [Hz]	CV of identified $f$	Damping ratio $\zeta$ [%]	CV of identified $\zeta$
1	0.0936	0.0030	0.68	0.7228
2	0.1384	0.0021	0.39	0.5095
3	0.3659	0.0021	0.34	0.4724
4	0.4238	0.0008	0.20	0.3520
5	0.4747	0.0007	0.12	0.8658
6	0.5055	0.0011	0.17	0.3027
7	0.5224	0.0008	0.19	0.4954
8	0.7953	0.0013	0.26	0.7801
9	0.9648	0.0015	0.27	0.4011
10	1.1505	0.0005	0.13	0.4609
11	1.1909	0.0005	0.12	0.6020
12	1.2507	0.0009	0.15	0.2338

the identified modal parameters with those already available results from other research groups, only the first 12 modes of the Canton Tower are of interest in this paper. To select the stable poles and construct the stabilization diagram, following tolerance values are used in Eq. (16)

$$f_{tol} = 1\%, \zeta_{tol} = 10\% \text{ and } MAC_{tol} = 0.99 \quad (16)$$

According to the modal frequency range provided by the reduced-order FE model of the Canton Tower (Lin *et al.* 2010), the first 12 modes are within 1.3 Hz, so the acceleration measurements are downsampled by a factor of 16 to decrease the computation effort. As an example, the stabilization diagram for the data set no. 61 is shown in Fig. 5.

The mean values and CV (Coefficient of Variation) values of the identified natural frequencies and damping ratios are listed in Table 1. As a verification, the results from the ARV technique are also compared with those obtained using the data-driven stochastic subspace identification (SSI-

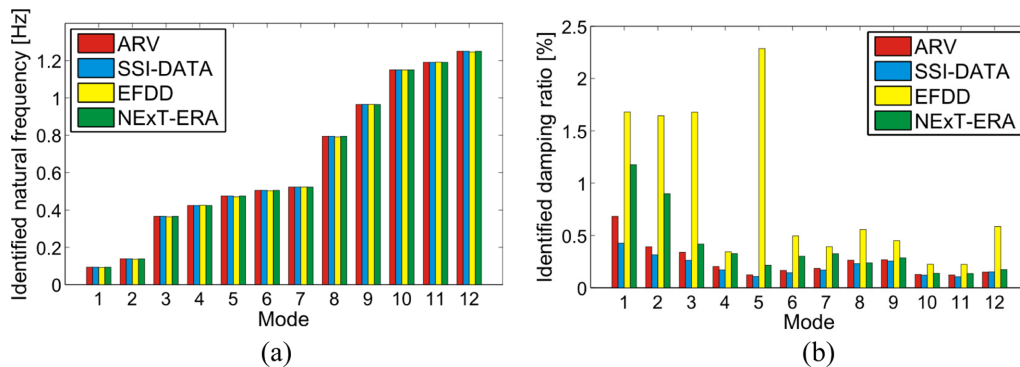


Fig. 6 Comparison of the results from the ARV, the SSI-DATA, the EFDD and the NExT-ERA: (a) Identified natural frequencies and (b) identified damping ratios

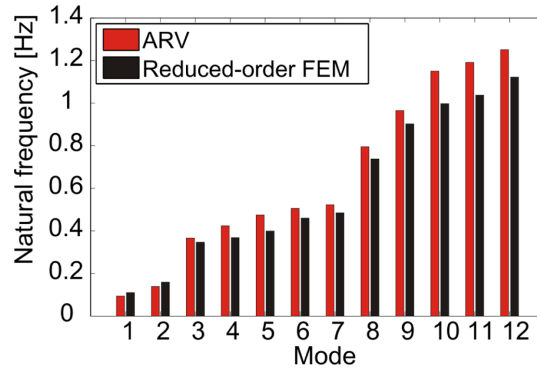


Fig. 7 Comparison of the identified natural frequencies with those of the FE model

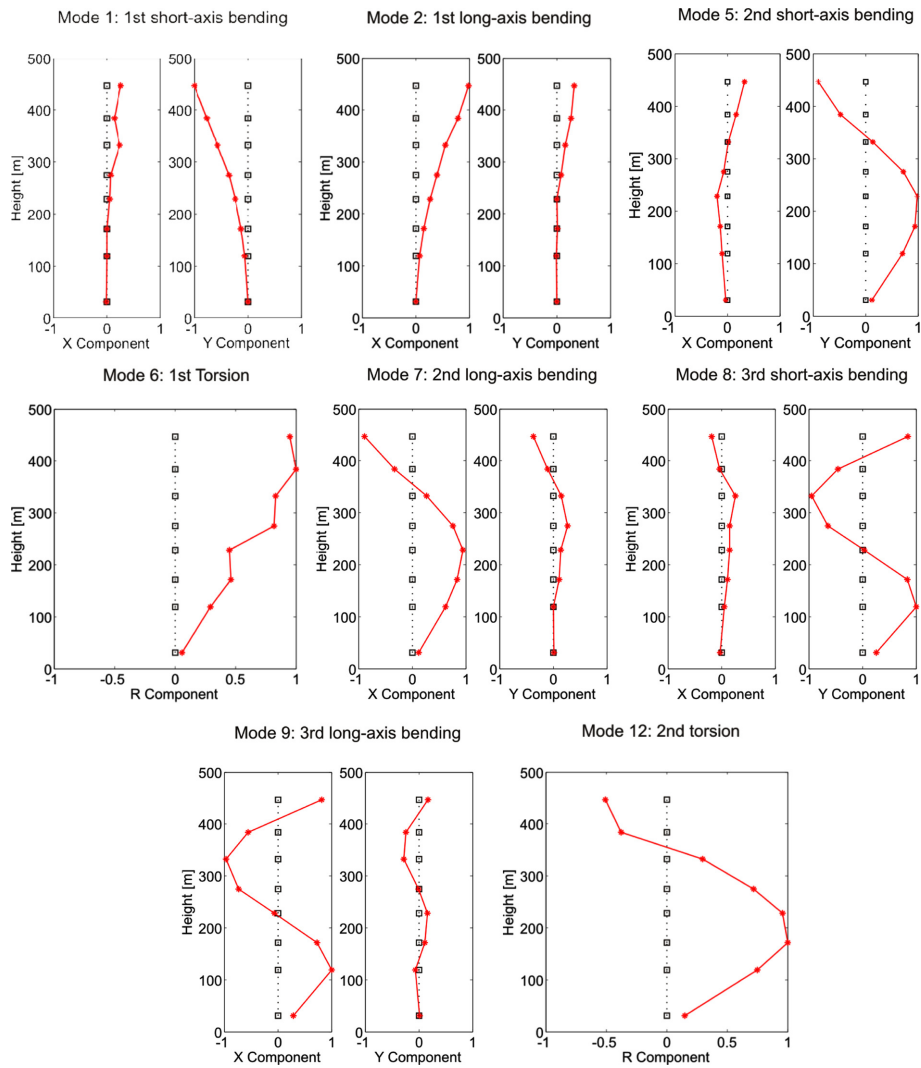


Fig. 8 The extracted mode shapes of the main tower (□ ·····: Ref.; \*—: Mode shape)

DATA) method (Faravelli *et al.* 2010a), the enhanced frequency domain decomposition (EFDD) algorithm (Ye *et al.* 2011), and an improved automatic modal identification method based on NExT-ERA (Ye *et al.* 2011). The comparison bar plots are shown in Fig. 6.

As shown in Fig. 6, the identified natural frequencies from these four methods have a quite good accordance with each other. However, big differences are noticed for the identified damping ratios, especially for the lower modes. Compared to the damping ratios from the EFDD and the NExT-ERA, the values from the ARV and the SSI-DATA show a relatively good agreement. Furthermore, Fig. 7 gives the comparison of the identified natural frequencies with those of the reduced-order FE model in presented in (Lin *et al.* 2010). It can be seen that there are still some differences, especially for modes 10, 11 and 12.

The identified mode shapes of the main tower, not including the antenna mast, are plotted in Fig. 8. It is observed that a sudden change appears in some lower bending modes, e.g., in the  $x$ -component of mode 1 and the  $x$ -component of mode 8. Such kind of sudden changes are also noticed in the identified mode shapes when using the SSI-DATA method implemented in the MATLAB toolbox MACEC and in the study results presented in (Loh *et al.* 2011, Ye *et al.* 2011). However, the mode shapes calculated from the tower model do not show such sudden changes (Lin *et al.* 2010). One explanation given to this is that the vibration modes are nonproportionally damped due to the special tube-in-tube geometry of the tower (Ye *et al.* 2011).

## 5. Environmental effects on the identified modal parameters

Another interest of this paper is to investigate the environmental effects on the identified modal parameters so that some reference information can be provided for later damage identification study.

The 1 h mean wind velocity and wind direction are plotted in Fig. 9, where 0 degree wind direction corresponds to the north. The corresponding wind rose diagram is given in Fig. 10. It can be seen that after 2 a.m. on January 20, 2010, the wind direction becomes relatively stable. Therefore, to reduce effects from varying wind directions on the modal parameters, only measurements after 2 a.m. on January 20, 2010 were used in the study between wind speed and modal parameters. The discussion on the correlation between the temperature and the modal parameters considers all measurement data. In order to eliminate the effects of outliers, robust regression analysis technique has been applied in the following study by using the MATLAB function *robustfit*.

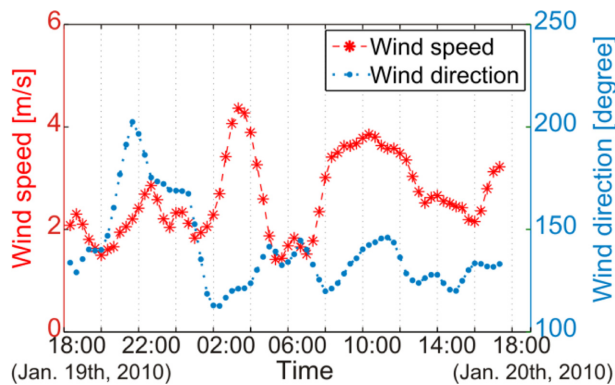


Fig. 9 1 h mean wind speed and wind direction

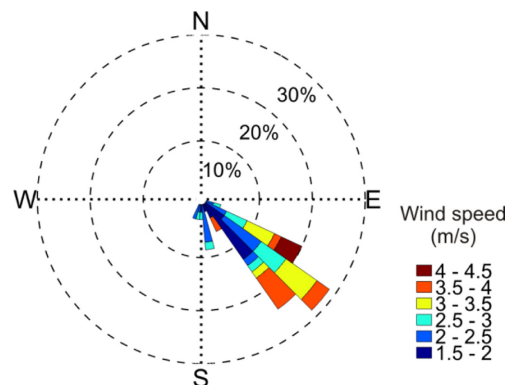


Fig. 10 Wind rose diagram

### 5.1 Natural frequency vs. temperature

Fig. 11 clearly reveals that the natural frequencies of the first 12 modes decrease as the temperature increases. This result agrees well with that presented in (Faravelli *et al.* 2010a). Assume the straight-line fit model has the form in Eq. (17)

$$f_i = \beta_0 + \beta_1 T \quad (17)$$

where  $f_i$  is the  $i$ -th natural frequency,  $T$  is the ambient temperature,  $\beta_0$  and  $\beta_1$  are the regression coefficients. The values of  $\beta_0$  and  $\beta_1$  for the first 12 modes are listed in Table 2.

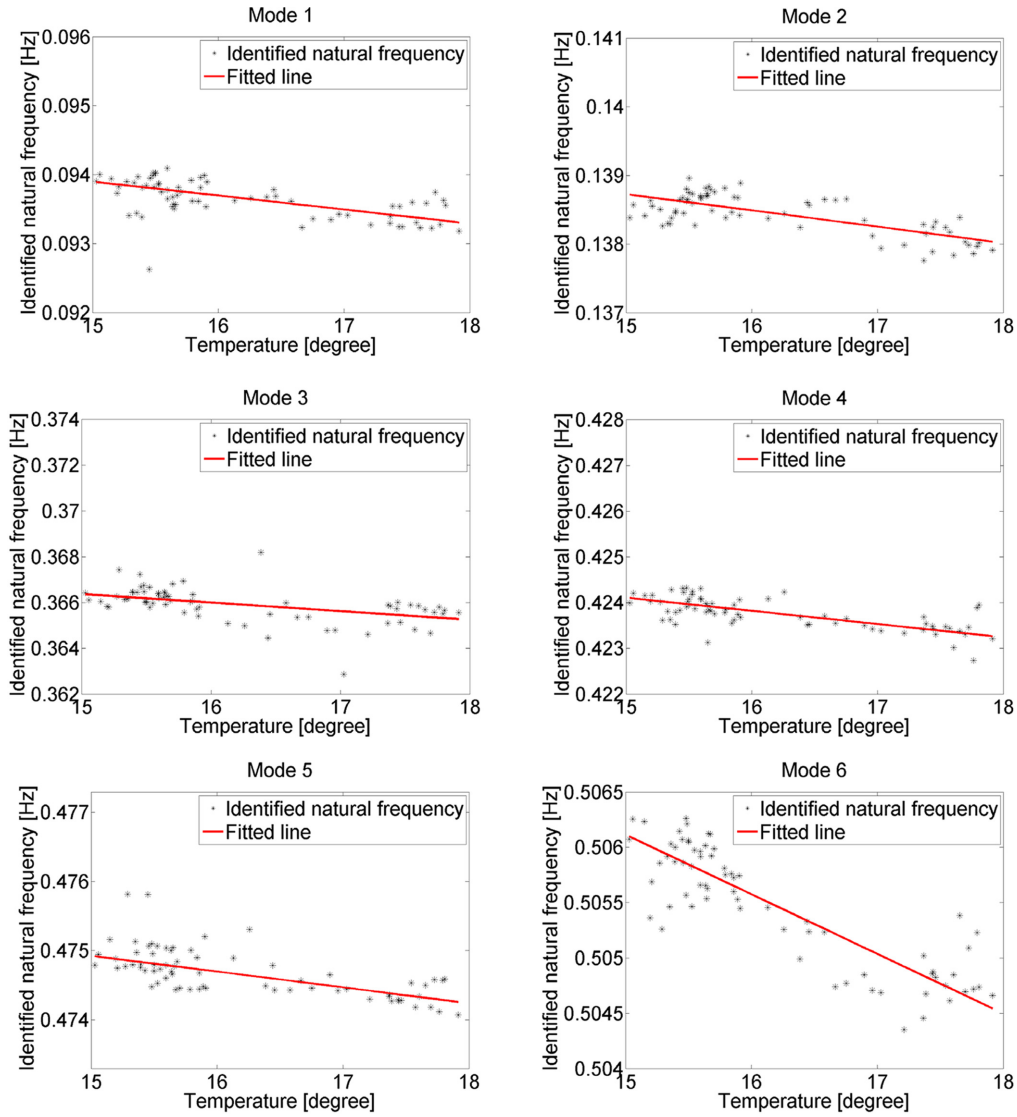


Fig. 11 Changes of natural frequencies with respect to the ambient temperature

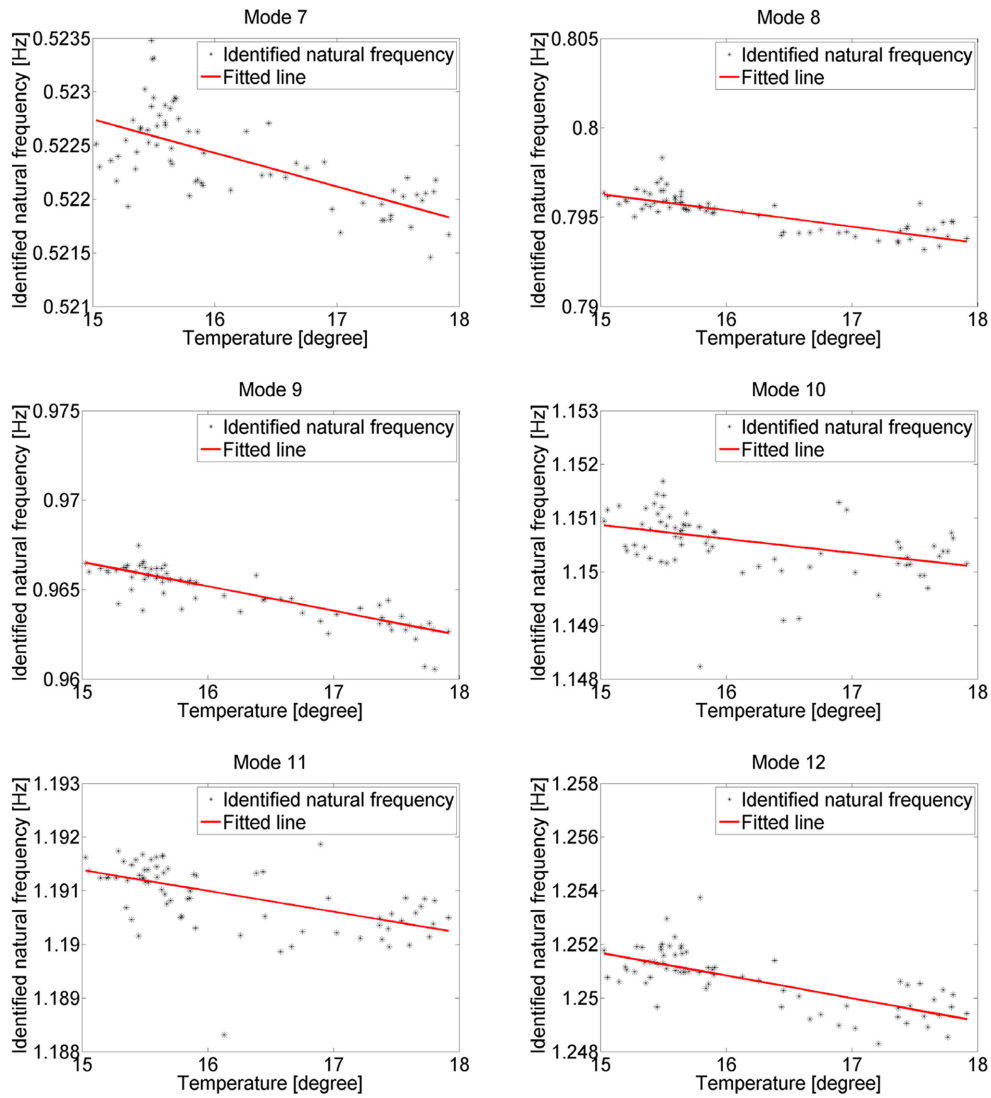


Fig. 11 Continued

Table 2 Correlation between the natural frequency and the ambient temperature

	Mode 1	Mode 2	Mode 3	Mode 4	Mode 5	Mode 6
$\beta_0$	0.0969	0.1423	0.3720	0.4284	0.4784	0.5142
$\beta_1 (10^{-4})$	-2.0285	-2.3705	-3.7537	-2.8865	-2.2902	-5.3893
	Mode 7	Mode 8	Mode 9	Mode 10	Mode 11	Mode 12
$\beta_0$	0.5274	0.8100	0.9871	1.1548	1.1972	1.2644
$\beta_1 (10^{-4})$	-3.1320	-9.1326	-0.0014	-2.6042	-3.8919	-8.4954

### 5.2 Natural frequency vs. wind speed

In Fig. 12, no uniform trend is reflected. The identified natural frequencies just slightly fluctuate around their mean values. Such fluctuation might be due to the temperature changes within the 24 hours. Similar results can also be found in (AlSaleh 2010), where the dynamic properties of the Canton Tower are studied under five typhoon events, and it is reported that the natural frequencies

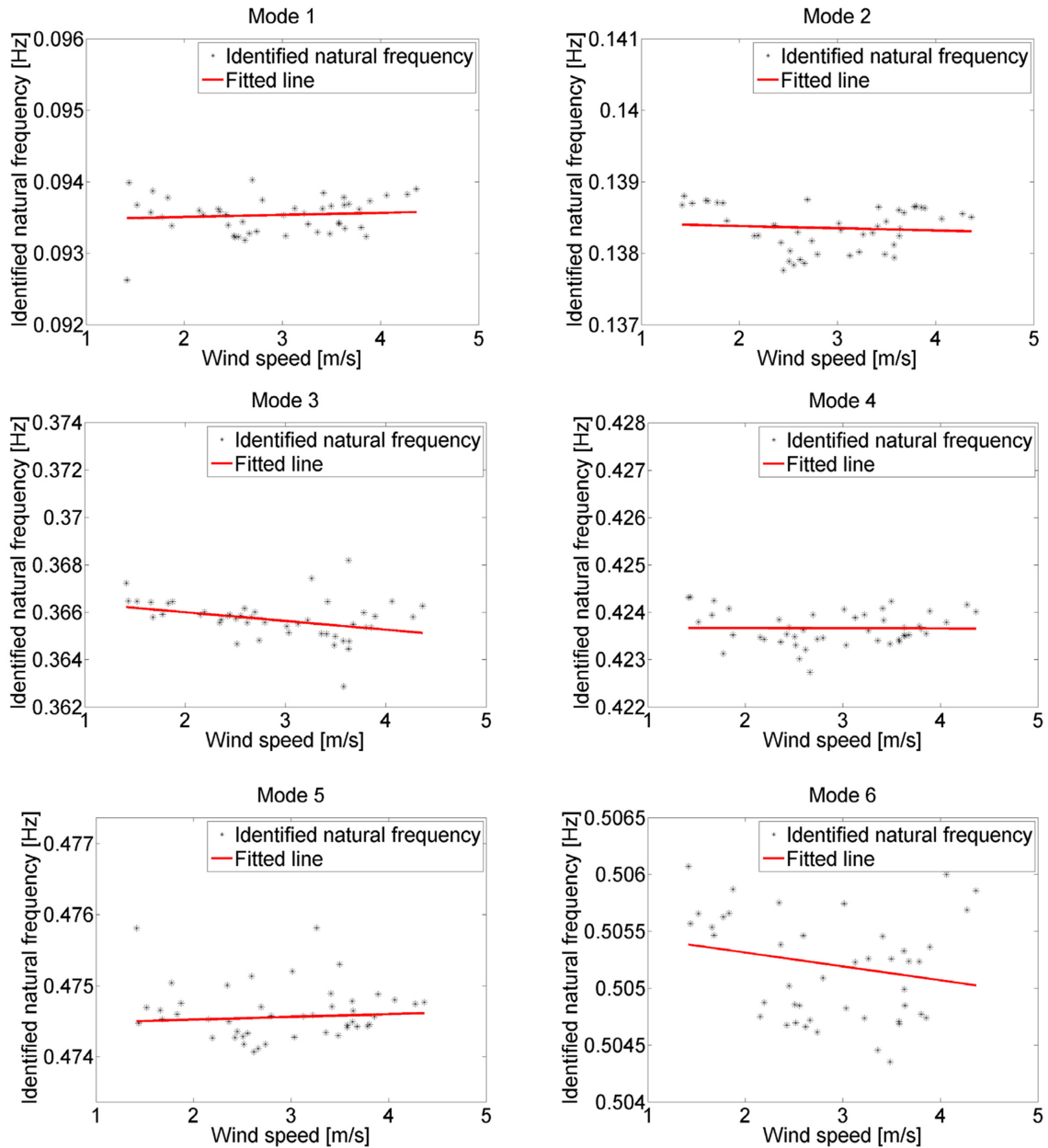


Fig. 12 Changes of the natural frequencies with respect to the wind speed

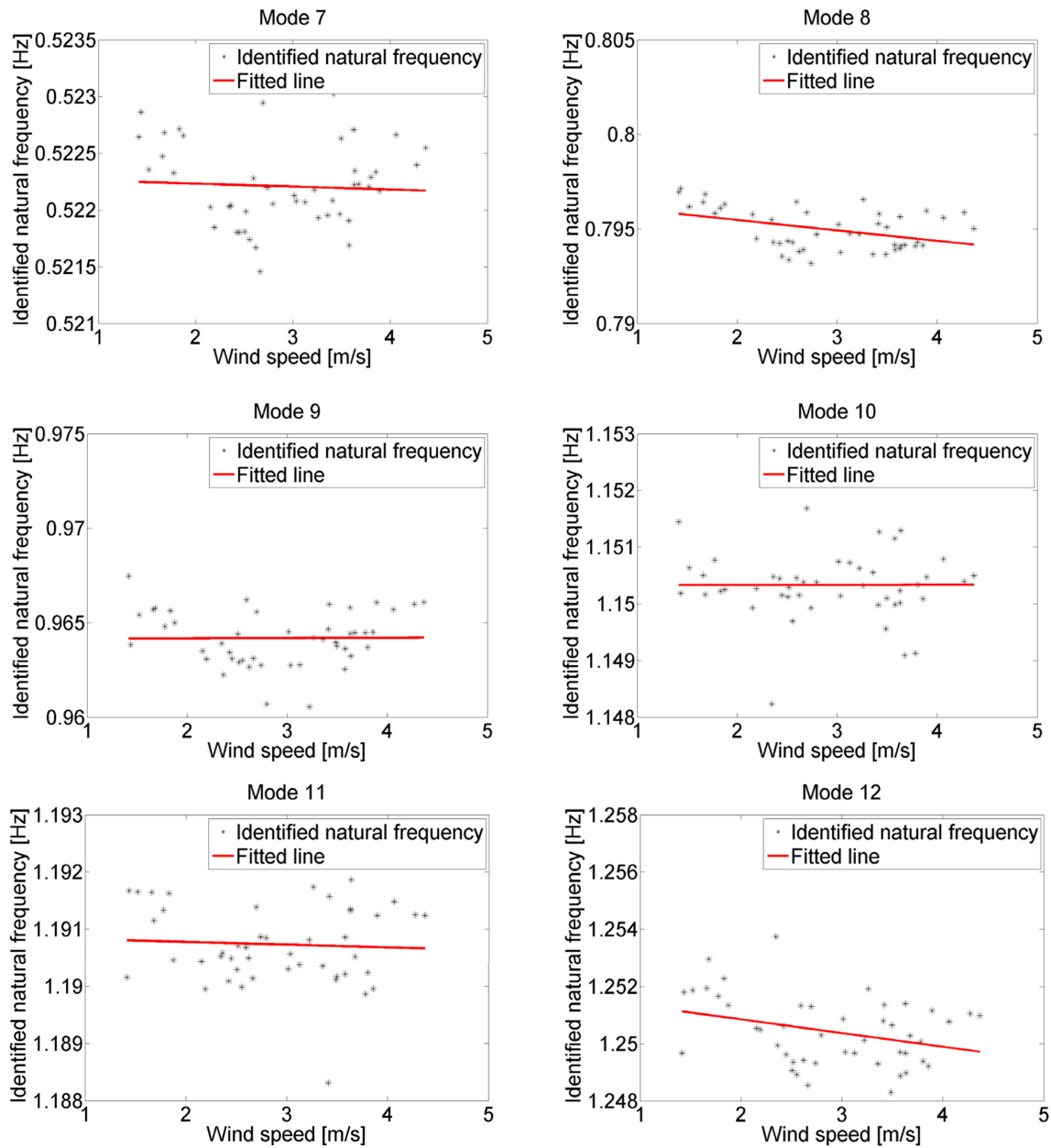


Fig. 12 Continued

of the tower under each typhoon also fluctuate slightly but do not follow a particular trend.

### 5.3 Damping ratio vs. temperature

In Fig. 13, the changes of identified damping ratios against the ambient temperatures are presented. No uniform trend can be observed.

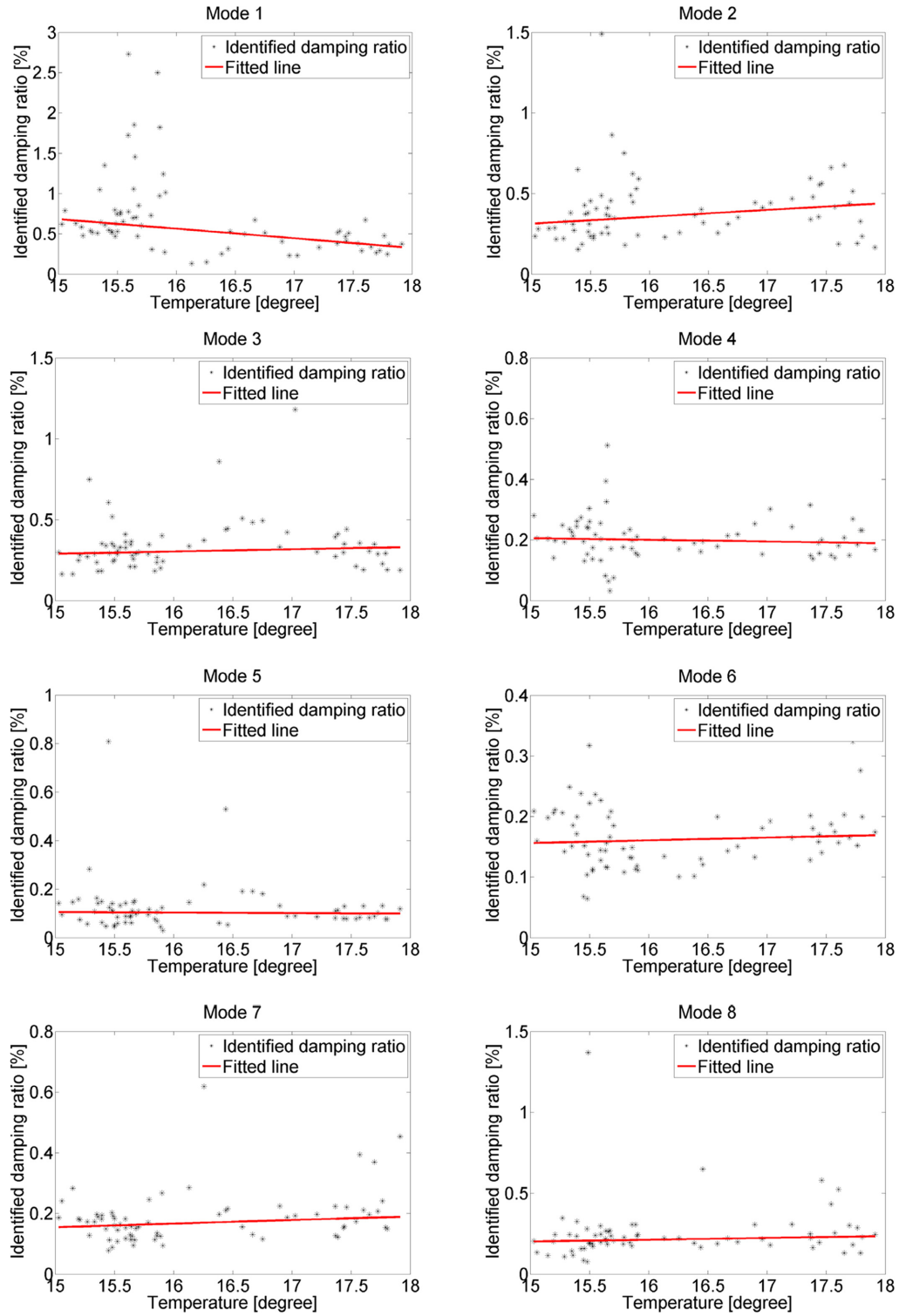


Fig. 13 Changes of the damping ratios with respect to the temperature

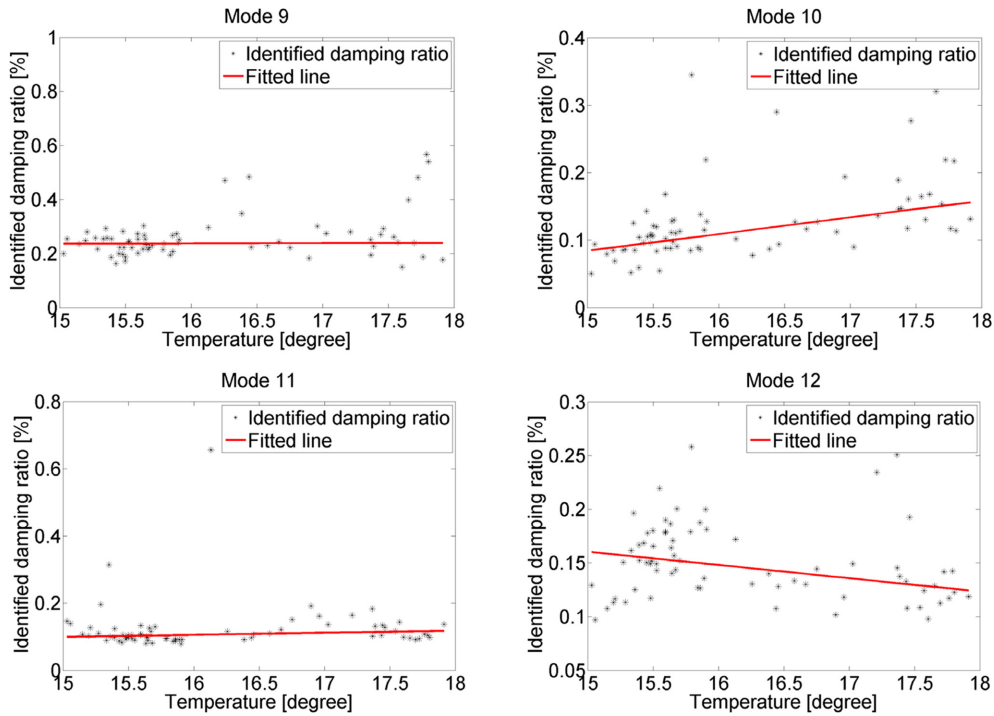


Fig. 13 Continued

#### 5.4 Damping ratio vs. wind speed

Fig. 14 does not show a uniform trend for this case. This result is actually in common with those in (Faravelli *et al.* 2010a, Ye *et al.* 2011) and it might be due to the low wind speed during the measurement (mean wind speed of around 3 m/s during one hour at the height of 461 m). For the case that the tower is subjected to high wind loads during a typhoon events, it is reported that the identified damping ratios increase with the wind speed (AlSaleh 2010), and the damping ratios obtained under typhoon are much higher than those extracted under normal ambient excitation condition (Guo *et al.* 2011).

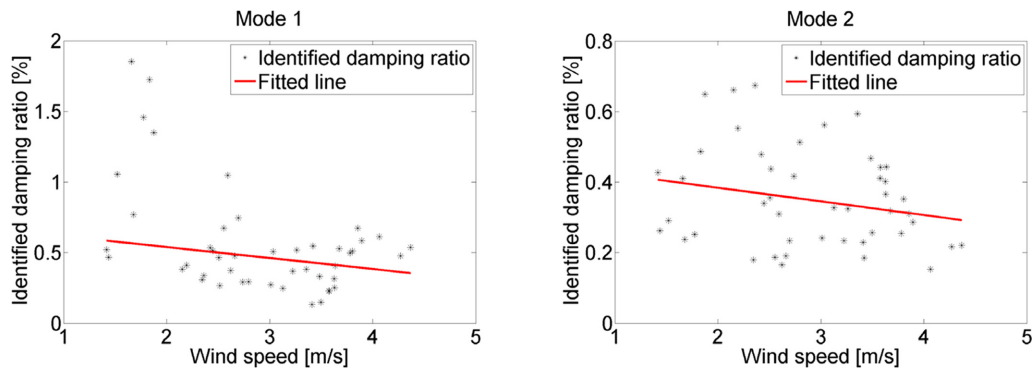


Fig. 14 Changes of the damping ratios with respect to the wind speed

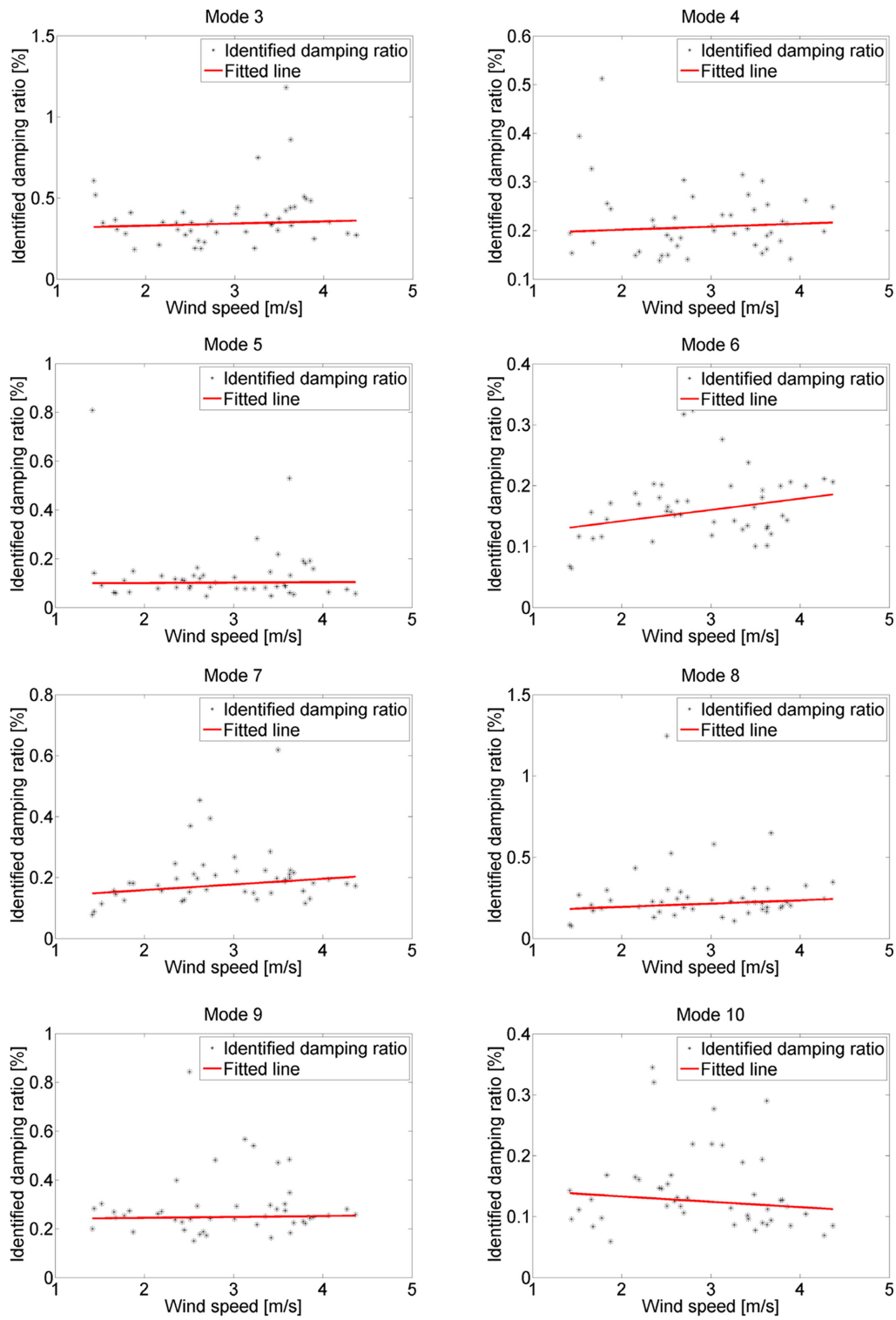


Fig. 14 Changes of the damping ratios with respect to the wind speed

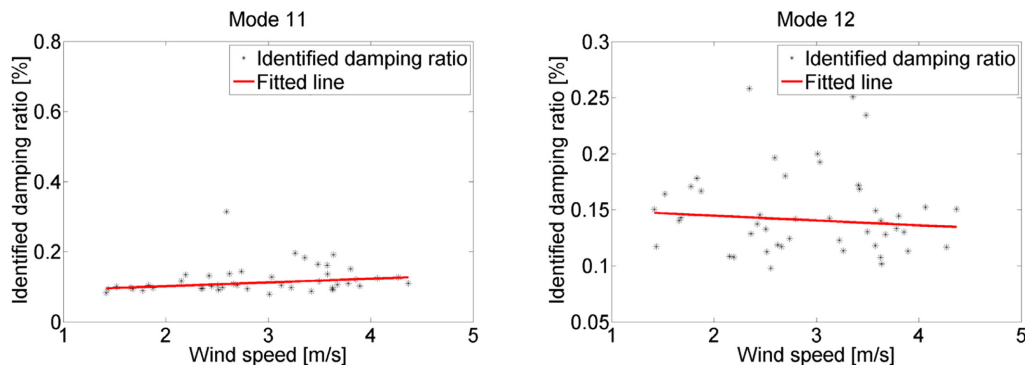


Fig. 14 Changes of the damping ratios with respect to the wind speed

## 6. Conclusions

In this contribution, the vector autoregressive models (ARV) method is adopted for the operational modal analysis (OMA) of the Canton Tower. The identified modal parameters (natural frequencies, damping ratios and mode shapes) are presented and compared with the available results, which are extracted by means of the data-driven stochastic subspace identification (SSI-DATA) algorithm, the enhanced frequency domain decomposition (EFDD) technique, and the NExT-ERA method. It can be concluded that the ARV method could also be considered as an efficient tool for OMA application. For the next step, the reduced-order FE model of the Canton tower will be updated according to the modal properties presented in this paper.

## Acknowledgements

The authors want to express their thanks to Prof. Yiqing Ni from the Hong Kong Polytechnic University for sharing the field measurement data from the SHM system of the Canton Tower. The work described in this paper was supported by the Multi Modal Sensor Systems for Environmental Exploration and Safety (MOSES) program in NRW, Germany, and by a grant from the Germany/Hong Kong Joint Research Scheme sponsored by the Research Grants Council of Hong Kong and the German Academic Exchange Service (Reference No. G\_HK035/11) .

## References

- Chen, W.H., Lu, Z.R., Chen, S.H., Ni, Y.Q. and Liao, W.Y. (2011a), "Monitoring dynamic characteristics for a supertall structure under different loading events", *Adv. Mater.*, **243-249**, (5356-5359).
- Chen, W.H., Lu, Z.R., Lin, W., Chen, S.H., Ni, Y.Q., Xia, Y. and Liao, W.Y. (2011b), "Theoretical and experimental modal analysis of the Guangzhou New TV Tower", *Eng. Struct.*, **33**(12), 3628-3646.
- Faravelli, L., Ubertini, F. and Fuggini, C. (2010a), "Subspace identification of the Guangzhou New TV Tower", *Proceedings of the 5th World Conference on Structural Control and Monitoring*, Shinjuku, Tokyo, July 12-14.
- Faravelli, L., Ubertini, F. and Fuggini, C. (2010b), "System identification toward FEM updating of a super high-rise building", *Proceedings of The 5th European Workshop on Structural Health Monitoring*, Sorrento, Naples, Italy,

- June 28-July 4.
- Farrar, C.R. and Worden, K. (2007), "An introduction to structural health monitoring", *Philos. T.R. Soc. A.*, **356**(1851), 303-315.
- Guo, Y.L., Ni, Y.Q. and Kareem, A. (2011), "Performance evaluation of the Guangzhou New TV Tower under winds based on full-scale monitoring data", *Proceedings of the 13th International Conference on Wind Engineering (ICWE 13)*, Amsterdam, Netherlands, July 10-15.
- Kraemer, P. and Fritzen, C.P. (2010a), "Aspects of operational modal analysis for structures of offshore wind energy plants", *Proceedings of The 28th International Modal Analysis Conference (IMAC XXIX)*, Jacksonville, Florida, USA, February 1-4.
- Kraemer, P. and Fritzen, C.P. (2010b), "Vibration analysis for structures of offshore wind energy plants", *Proceedings of the 10th German Wind Energy Conference (DEWEK 2010)*, Bremen, Germany, November 17-18.
- Lin, W., Ni, Y.Q., Xia, Y. and Chen, W.H. (2010), "Field Measurement data and a reduced-order finite element model for Task I of the SHM benchmark problem for high-rise structures", *Proceedings of the 5th World Conference on Structural Control and Monitoring*, Shinjuku, Tokyo, July 12-14.
- Loh, C.H., Liu, Y.C. and Ni, Y.Q. (2011), "SSA-Based stochastic subspace identification of structures from output-only vibration measurements", *Proceedings of the 8th International Workshop on Structural Health Monitoring (IWSHM 2011)*, Stanford, CA, USA, September 13-15.
- Neumaier, A. and Schneider, T. (2001), "Estimation of parameters and eigenmodes of multivariate autoregressive models", *ACM T. Math. Software*, **27**(1), 27-57.
- Ni, Y.Q., Wong, K.Y. and Xia, Y. (2011), "Health checks through landmark bridges to sky-high structures", *Adv. Struct. Eng.*, **14**(1), 103-119.
- Ni, Y.Q., Xia, Y., Liao, W.Y. and Ko, J.M. (2009), "Technology innovation in developing the structural health monitoring system for Guangzhou New TV Tower", *Struct. Control Health Monit.*, **16**(1), 73-98.
- Ni, Y.Q., Xia, Y., Lin, W., Chen, W.H. and Ko, J.M. (2012), "SHM benchmark for high-rise structures: a reduced-order finite element model and field measurement data", *Smart Struct. Syst.*, in this issue.
- Ni, Y.Q. and Zhou, H.F. (2010), "Guangzhou New TV Tower: integrated structural health monitoring and vibration control", *Proceedings of the 2010 Structures Congress*, Orlando, Florida, USA, May 12-15.
- Niu, Y., Klinkov, M. and Fritzen, C.P. (2011a), "Extension of the generalized unknown input kalman filter for online-reconstruction of external structural loads", *Proceedings of the 8th International Workshop on Structural Health Monitoring (IWSHM 2011)*, Stanford, CA, USA, September 13-15.
- Niu, Y., Kraemer, P. and Fritzen, C.P. (2011b), "Operational modal analysis for a benchmark high-rise structure", *Proceedings of the 8th International Workshop on Structural Health Monitoring (IWSHM 2011)*, Stanford, CA, USA, September 13-15.
- Niu, Y., Kraemer, P. and Fritzen, C.P. (2011c), "Operational modal analysis for the Guangzhou New TV Tower", *Proceedings of the 29th International Modal Analysis Conference (IMAC XXIX)*, Jacksonville, Florida, USA, January 28-31.
- Schneider, T. and Neumaier, A. (2001), "Algorithm 808: ARfit | A Matlab package for the estimation of parameters and eigenmodes of multivariate autoregressive models", *ACM T. Math. Software*, **27**(1), 58-65.
- Xia, Y., Ni, Y.Q., Ko, J.M., Liao, W.Y. and Chen, W.H. (2009), "ANCRiSST benchmark problem on structural health monitoring of high-rise slender structures - Phase I: field vibration measurement", *Proceedings of the 5th International Workshop on Advanced Smart Materials and Smart Structures Technology*, Boston, MA, USA, July 29-August 1.
- Ye, X., Yan, Q., Wang, W., Yu, X. and Zhu, T. (2011), "Output-only modal identification of Guangzhou New TV Tower subject to different environment effects", *Proceedings of the 6th International Workshop on Advanced Smart Materials and Smart Structures Technology*, ANCRiSST2011, Dalian, China, July 25-26.
- Zhang, L., Brincker, R. and Andersen, P. (2005), "An overview of operational modal analysis: major development and issues", *Proceedings of the 1st International Operational Modal Analysis Conference*, Copenhagen, Denmark, April 26-27.