

Vibrations and acoustics

7. Experimental Modal Analysis

Arnaud Deraemaeker (adema@ulb.ac.be)

1

Outline of the chapter

- *Principle of EMA
- *Measuring FRFs
- *SDOF Identification
- *MDOF Identification

2

Principle of EMA



1. Measure FRFs
2. Estimate poles (natural frequencies)
3. Identify mode shapes

3

Principle of EMA

Impact Testing for Modal Analysis

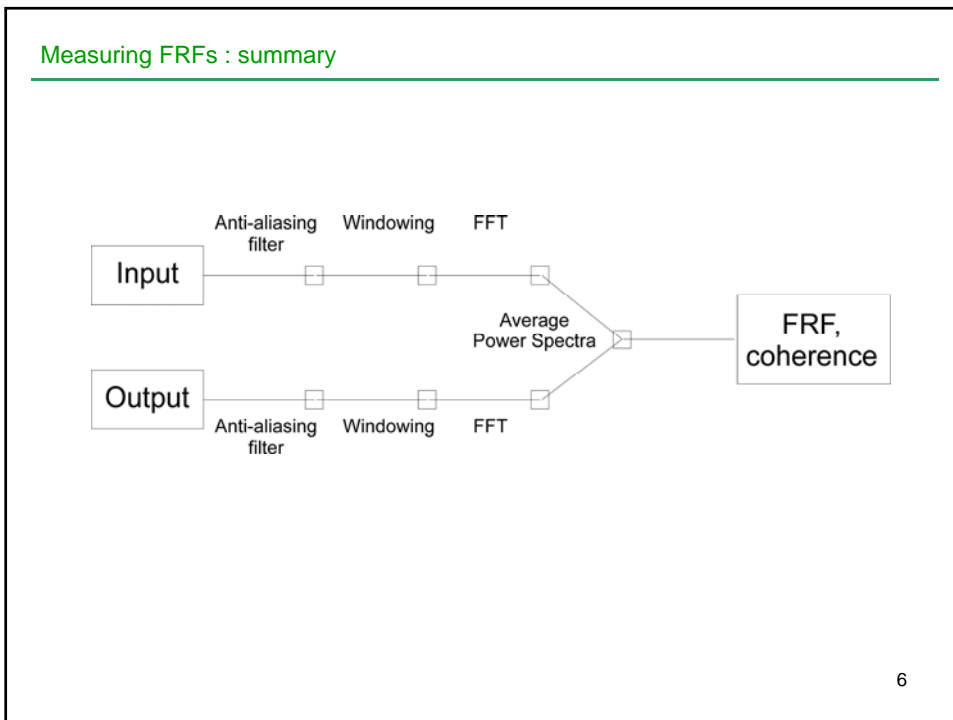
DYTRAN INSTRUMENTS, INC. Vibrant TECHNOLOGY

4

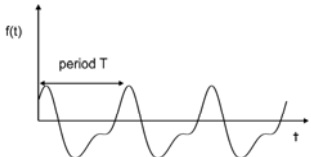

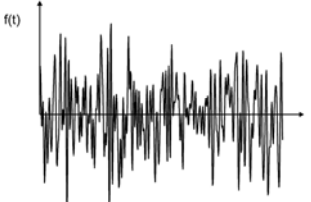


Measuring FRFs

5



Use of windows : summary

	<p>Periodic</p>	<p>No window (synchronisation !)</p>
	<p>Transient</p>	<p>Exponential window to reduce effect of noise (output) + force window (input)</p>
	<p>Random</p>	<p>Hanning window to reduce leakage</p>

7

Measuring FRFs

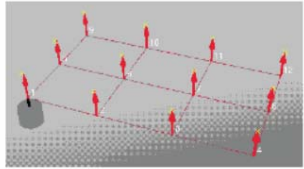
Shaker excitation

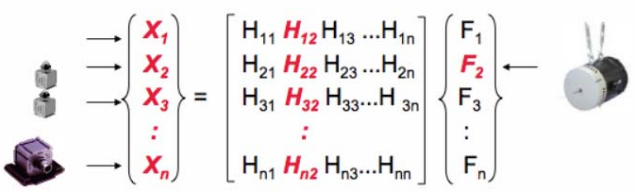
One column

- Single Fixed Excitation (reference)
- Single Roving Response **SISO**

or

- Multiple (Roving) Responses **SIMO**
- Multiple-Output: Optimize **data consistency**





$$\begin{matrix} \text{Structure} \\ \rightarrow \\ \begin{pmatrix} X_1 \\ X_2 \\ X_3 \\ \vdots \\ X_n \end{pmatrix} \end{matrix} = \begin{bmatrix} H_{11} & H_{12} & H_{13} & \dots & H_{1n} \\ H_{21} & H_{22} & H_{23} & \dots & H_{2n} \\ H_{31} & H_{32} & H_{33} & \dots & H_{3n} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ H_{n1} & H_{n2} & H_{n3} & \dots & H_{nn} \end{bmatrix} \begin{matrix} \leftarrow \text{Shaker} \\ \begin{pmatrix} F_1 \\ F_2 \\ F_3 \\ \vdots \\ F_n \end{pmatrix} \end{matrix}$$

8

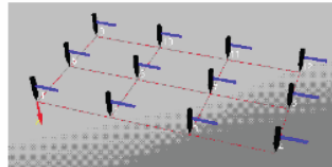
Measuring FRFs

Roving hammer excitation

One row

- One Roving Excitation
- One Fixed Response (reference)

SISO



$$\begin{matrix} \text{hammer} \end{matrix} \rightarrow \begin{Bmatrix} X_1 \\ X_2 \\ X_3 \\ \vdots \\ X_n \end{Bmatrix} = \begin{bmatrix} H_{11} & H_{12} & H_{13} & \dots & H_{1n} \\ H_{21} & H_{22} & H_{23} & \dots & H_{2n} \\ H_{31} & H_{32} & H_{33} & \dots & H_{3n} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ H_{n1} & H_{n2} & H_{n3} & \dots & H_{nn} \end{bmatrix} \begin{Bmatrix} F_1 \\ F_2 \\ F_3 \\ \vdots \\ F_n \end{Bmatrix} \leftarrow \begin{matrix} \text{hammer} \end{matrix}$$

9

Measuring FRFs

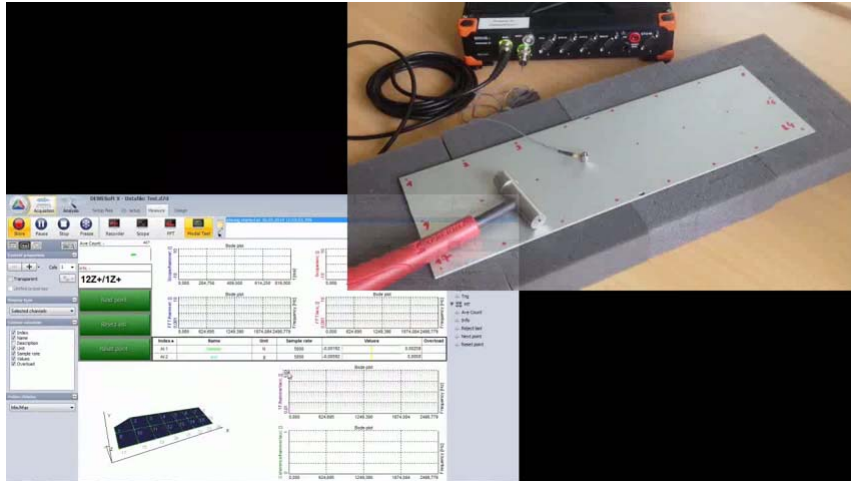
Reciprocity

$$\begin{matrix} \text{hammer} \\ \text{hammer} \\ \text{hammer} \\ \vdots \\ \text{hammer} \end{matrix} \rightarrow \begin{Bmatrix} X_1 \\ X_2 \\ X_3 \\ \vdots \\ X_n \end{Bmatrix} = \begin{bmatrix} H_{11} & H_{12} & H_{13} & \dots & H_{1n} \\ H_{21} & H_{22} & H_{23} & \dots & H_{2n} \\ H_{31} & H_{32} & H_{33} & \dots & H_{3n} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ H_{n1} & H_{n2} & H_{n3} & \dots & H_{nn} \end{bmatrix} \begin{Bmatrix} F_1 \\ F_2 \\ F_3 \\ \vdots \\ F_n \end{Bmatrix} \leftarrow \begin{matrix} \text{hammer} \end{matrix}$$

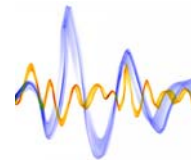
$$\begin{matrix} \text{hammer} \end{matrix} \rightarrow \begin{Bmatrix} X_1 \\ X_2 \\ X_3 \\ \vdots \\ X_n \end{Bmatrix} = \begin{bmatrix} H_{11} & H_{12} & H_{13} & \dots & H_{1n} \\ H_{21} & H_{22} & H_{23} & \dots & H_{2n} \\ H_{31} & H_{32} & H_{33} & \dots & H_{3n} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ H_{n1} & H_{n2} & H_{n3} & \dots & H_{nn} \end{bmatrix} \begin{Bmatrix} F_1 \\ F_2 \\ F_3 \\ \vdots \\ F_n \end{Bmatrix} \leftarrow \begin{matrix} \text{hammer} \end{matrix}$$

10

Measuring FRFs in practice



11



SDOF identification

12

The pole-residue model in the frequency domain

The frequency response function of a one dof system is :

$$H(\omega) = \frac{X(\omega)}{F(\omega)} = \frac{1}{m} \frac{1}{\omega_n^2 - \omega^2 + 2j\xi\omega\omega_n}$$

The Pole-residue model in the frequency domain is :

$$H(\omega) = \frac{R}{j\omega - \lambda} + \frac{R^*}{j\omega - \lambda^*}$$

→

$$\lambda = -\xi\omega_n + j\omega_d$$

$$R = -\frac{j}{2m\omega_d}$$

$$\omega_d = \omega_n \sqrt{1 - \xi^2}$$

13

The pole-residue model in the time domain

The impulse response function of a one DOF system is :

$$h(t) = \frac{e^{-\xi\omega_n t}}{m\omega_d} \sin(\omega_d t)$$

The Pole-residue model in the time domain is :

$$h(t) = R e^{\lambda t} + R^* e^{\lambda^* t}$$

→

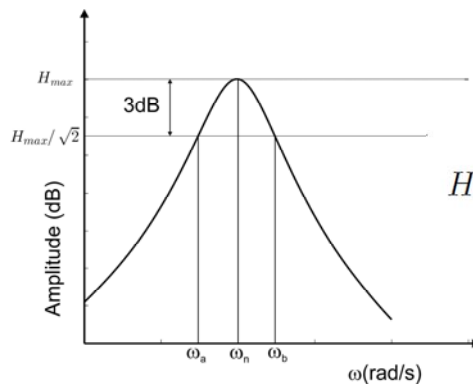
$$\lambda = -\xi\omega_n + j\omega_d$$

$$R = -\frac{j}{2m\omega_d}$$

$$\omega_d = \omega_n \sqrt{1 - \xi^2}$$

14

Estimating parameters : Peak-picking method



$$H(\omega) = \frac{1}{m} \frac{1}{\omega_n^2 - \omega^2 + 2j\xi\omega\omega_n}$$

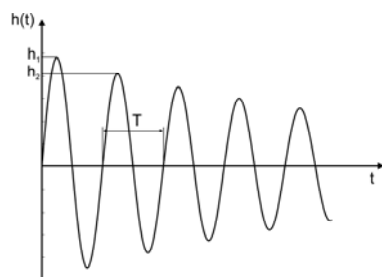
ω_n Natural frequency (ratio k/m)

$2\xi \simeq \frac{\omega_b - \omega_a}{\omega_n}$ Damping coefficient (b)

$m = \frac{1}{2\xi\omega_n^2 H_{max}}$ Mass (m)

15

Estimating parameters : logarithmic decrement method



$$h(t) = \frac{e^{-\xi\omega_n t}}{m\omega_d} \sin(\omega_d t)$$

$\xi = \frac{1}{2\pi N} \ln \left(\frac{h(t)}{h(t + NT)} \right)$ Damping coefficient (b)

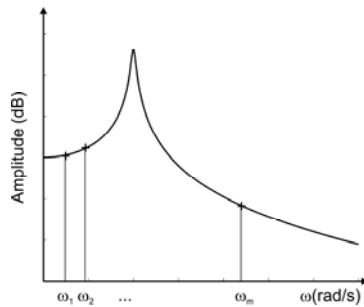
$\omega_d = \omega_n \sqrt{1 - \xi^2} = \frac{2\pi}{T}$ Natural frequency (ratio k/m)

$m = \frac{e^{-\xi\omega_n \frac{T}{2}}}{\omega_d h_1}$ Mass (m)

16

Estimating parameters : curve fitting

$$(k - \omega^2 m + j\omega b)X(\omega) = F(\omega)$$



$$\begin{bmatrix} -\omega_1^2 X(\omega_1) & j\omega_1 X(\omega_1) & X(\omega_1) \\ -\omega_2^2 X(\omega_2) & j\omega_2 X(\omega_2) & X(\omega_2) \\ \dots & \dots & \dots \\ -\omega_m^2 X(\omega_m) & j\omega_m X(\omega_m) & X(\omega_m) \end{bmatrix} \begin{Bmatrix} m \\ b \\ k \end{Bmatrix} = \begin{Bmatrix} F(\omega_1) \\ F(\omega_2) \\ \dots \\ F(\omega_m) \end{Bmatrix}$$

$$[A] \{\theta\} = \{F\} \quad \text{More equations than unknowns}$$

17

Estimating parameters : curve fitting

$$[A] \{\theta\} = \{F\} \quad \text{More equations than unknowns}$$

$$[A]^* [A] \{\theta\} = [A]^* \{F\}$$

$$\{\theta\} = ([A]^* [A])^{-1} [A]^* \{F\} = A^\dagger \{F\}$$

$$A^\dagger = ([A]^* [A])^{-1} [A]^*$$

Moore-Penrose pseudo-inverse : Least-squares solution

18



MDOF identification

19

The pole-residue model in the frequency domain

The FRF matrix of a MDOF system is :

$$X(\omega) = H(\omega)F(\omega) \quad H(\omega) = \sum_{i=1}^N \frac{\psi_i^T \psi_i}{\mu_i(\omega_i^2 + 2j\xi\omega\omega_i - \omega^2)}$$

The Pole-residue model in the frequency domain is :

$$H(\omega) = \sum_{i=1}^N \frac{R_i}{j\omega - \lambda_i} + \frac{R_i^*}{j\omega - \lambda_i^*} \quad R_i \text{ is a matrix}$$

$$\begin{aligned} \lambda_i &= -\xi_i\omega_i + j\omega_i\sqrt{1 - \xi_i^2} \\ R_i &= \frac{-j\psi_i^T \psi_i}{2\mu_i\omega_i\sqrt{1 - \xi_i^2}} \end{aligned}$$

20

The pole-residue model in the time domain

$$H(\omega) = \sum_{i=1}^N \frac{R_i}{j\omega - \lambda_i} + \frac{R_i^*}{j\omega - \lambda_i^*}$$

Inverse Fourier transform

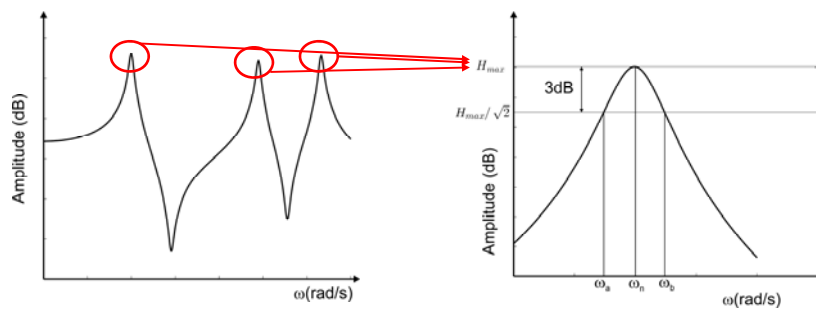
The Pole-residue model in the time domain is :

$$h(t) = \sum_{i=1}^N R_i e^{\lambda_i t} + R_i^* e^{\lambda_i^* t}$$

R_i is a matrix

21

Estimating the poles : peak picking method



$\omega_i, \xi_i \quad i = 1..N$ Natural frequencies and modal damping

22

Estimating the mode shapes : curve fitting in the frequency domain

$$\omega_i, \xi_i \quad i = 1..N \quad \longrightarrow \quad \lambda_i = -\xi_i \omega_i + j \omega_i \sqrt{1 - \xi_i^2}$$

For each single FRF :

$$H_{kl}(\omega) = \sum_{i=1}^N \frac{R_{i,kl}}{j\omega - \lambda_i} + \frac{R_{i,kl}^*}{j\omega - \lambda_i^*} \quad \text{l=input, k=output}$$

$$\begin{bmatrix} \frac{1}{j\omega_1 - \lambda_1} & \frac{1}{j\omega_1 - \lambda_1^*} & \frac{1}{j\omega_1 - \lambda_2} & \frac{1}{j\omega_1 - \lambda_2^*} & \dots & \dots & \frac{1}{j\omega_1 - \lambda_N} & \frac{1}{j\omega_1 - \lambda_N^*} \\ \frac{1}{j\omega_2 - \lambda_1} & \frac{1}{j\omega_2 - \lambda_1^*} & \frac{1}{j\omega_2 - \lambda_2} & \frac{1}{j\omega_2 - \lambda_2^*} & \dots & \dots & \frac{1}{j\omega_2 - \lambda_N} & \frac{1}{j\omega_2 - \lambda_N^*} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \frac{1}{j\omega_m - \lambda_1} & \frac{1}{j\omega_m - \lambda_1^*} & \frac{1}{j\omega_m - \lambda_2} & \frac{1}{j\omega_m - \lambda_2^*} & \dots & \dots & \frac{1}{j\omega_m - \lambda_N} & \frac{1}{j\omega_m - \lambda_N^*} \end{bmatrix} \begin{Bmatrix} R_{1,kl} \\ R_{1,kl}^* \\ R_{2,kl} \\ R_{2,kl}^* \\ \dots \\ \dots \\ R_{N,kl} \\ R_{N,kl}^* \end{Bmatrix} = \begin{Bmatrix} H_{kl}(\omega_1) \\ H_{kl}(\omega_2) \\ \dots \\ H_{kl}(\omega_m) \end{Bmatrix}$$

23

Estimating the mode shapes

$$[A] \{\theta\} = \{F\} \quad \longrightarrow \quad \{\theta\} = ([A]^* [A])^{-1} [A]^* \{F\} = A^\dagger \{F\}$$

$$R_{i,kl} = \frac{-j\psi_{i,k}\psi_{i,l}}{2\mu_i\omega_i\sqrt{1-\xi_i^2}}$$

Keep l fixed, vary k

$$\psi_{i,k} \quad \text{Mode shapes}$$

24

Estimating natural frequencies and mode shapes : the complex exponential method

$$h(t) = \sum_{i=1}^N R_i e^{\lambda_i t} + R_i^* e^{\lambda_i^* t}$$

$$h(t)_{kl} = \sum_{i=1}^N R_{i,kl} e^{\lambda_i t} + R_{i,kl}^* e^{\lambda_i^* t} \quad \text{l=input, k=output}$$

$$\downarrow$$

$$h(t)_{kl} = \sum_{i=1}^{2N} R_{i,kl} e^{\lambda_i t} \quad \left| \begin{array}{l} R_{i+N,kl} = R_{i,kl}^* \\ \lambda_{i+N} = \lambda_i^* \end{array} \right.$$

25

Estimating natural frequencies and mode shapes : the complex exponential method

$$\begin{array}{ll} h_0 = h(0)_{kl} = \sum_{i=1}^{2N} R_{i,kl} & h_0 = \sum_{i=1}^{2N} R_{i,kl} \\ h_1 = h(\Delta t)_{kl} = \sum_{i=1}^{2N} R_{i,kl} e^{\lambda_i(\Delta t)} & \xrightarrow{V_i = e^{\lambda_i(\Delta t)}} h_1 = \sum_{i=1}^{2N} R_{i,kl} V_i \\ \dots & \dots \\ h_L = h(L\Delta t)_{kl} = \sum_{i=1}^{2N} R_{i,kl} e^{\lambda_i(L\Delta t)} & h_L = \sum_{i=1}^{2N} R_{i,kl} V_i^L \end{array}$$

Because the roots appear as complex conjugate pairs, so do the V_i , therefore there exists real coefficients β such that :

$$\beta_0 + \beta_1 V_i + \beta_2 V_i^2 + \dots + \beta_L V_i^L = 0$$

$$\longrightarrow \sum_{j=0}^L \beta_j h_j = \sum_{i=1}^{2N} \left(R_{i,kl} \sum_{j=0}^L \beta_j V_i^j \right) = 0 \quad \forall V_i$$

26

Estimating natural frequencies and mode shapes : the complex exponential method

$$\sum_{j=0}^L \beta_j h_j = 0$$

$$\longrightarrow \sum_{j=0}^L \beta_j h_j V_i^n = \sum_{j=0}^L \beta_j h_{j+n} = 0$$

$$\begin{matrix} \downarrow \\ L = 2N \\ \beta_{2N} = 1 \end{matrix}$$

$$\begin{bmatrix} h_0 & h_1 & h_2 & \dots & h_{2N-1} \\ h_1 & h_2 & h_3 & \dots & h_{2N} \\ \dots & \dots & \dots & \dots & \dots \\ h_{2N-1} & h_{2N} & h_{2N+1} & \dots & h_{4N-2} \end{bmatrix} \begin{Bmatrix} \beta_0 \\ \beta_1 \\ \dots \\ \beta_{2N-1} \end{Bmatrix} = - \begin{Bmatrix} h_{2N} \\ h_{2N+1} \\ \dots \\ h_{4N-1} \end{Bmatrix}$$

$$\beta_0 + \beta_1 V_i + \beta_2 V_i^2 + \dots + \beta_L V_i^L = 0 \longrightarrow V_i = e^{\lambda_i(\Delta t)} \longrightarrow \lambda_i = \frac{\ln(V_i)}{\Delta t}$$

27

Estimating natural frequencies and mode shapes : the complex exponential method

$$\begin{aligned} h_0 &= \sum_{i=1}^{2N} R_{i,kl} \\ h_1 &= \sum_{i=1}^{2N} R_{i,kl} V_i \\ \dots & \dots \\ h_L &= \sum_{i=1}^{2N} R_{i,kl} V_i^L \end{aligned} \longrightarrow \begin{bmatrix} 1 & 1 & \dots & 1 \\ V_1 & V_2 & \dots & V_{2N} \\ V_1^2 & V_2^2 & \dots & V_{2N}^2 \\ \dots & \dots & \dots & \dots \\ V_1^{2N-1} & V_2^{2N-1} & \dots & V_{2N}^{2N-1} \end{bmatrix} \begin{Bmatrix} R_{1,kl} \\ R_{2,kl} \\ R_{3,kl} \\ \dots \\ R_{2N,kl} \end{Bmatrix} = \begin{Bmatrix} h_0 \\ h_1 \\ h_2 \\ \dots \\ h_{2N-1} \end{Bmatrix}$$

$$R_{i,kl} = \frac{-j\psi_{i,k}\psi_{i,l}}{2\mu_i\omega_i\sqrt{1-\xi_i^2}}$$

$$\downarrow$$

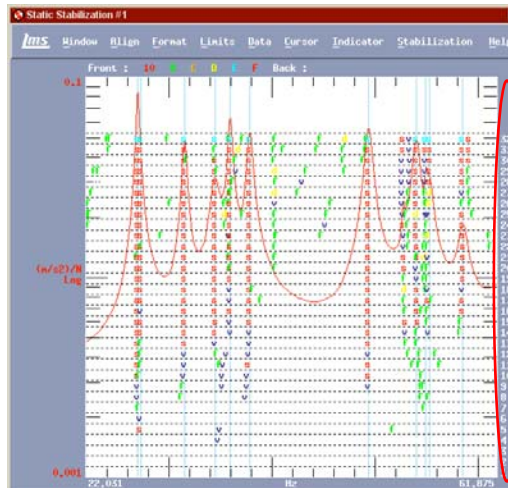
$$\psi_{i,k}$$

28

Model order and stabilisation diagram

$$h(t)_{kl} = \sum_{i=1}^{2N} R_{i,kl} e^{\lambda_i t} \quad \longrightarrow \quad N ?$$

- s: stable in both freq and damping
- d: stable in damping
- f: stable in frequency



N

29