

goals of EOF analysis

- 1. economy of description → reduce data ^{reveal} → obvious redundancies
- 2. orthogonalization → transform into a set of variables that are mutually uncorrelated, linear independent

2009 "When a large number of measurements are available, it is natural to enquire whether they could be replaced by a fewer number of the measurements or of their functions, without loss of much information, for convenience in the analysis and in the interpretation of data" Rao (1964)
 great Indian statistician

vector space A (measured variables)

vector space B (statistical variables)

currents @ x_1	=	$u(t, \vec{x}_1)$	=	$a_{11}(\vec{x}_1)$	• $Z_1(t)$ +	$a_{21}(x_1)$
currents @ x_2		$u(t, \vec{x}_2)$		$a_{12}(\vec{x}_2)$		$a_{22}(x_2)$
currents @ x_3		$u(t, \vec{x}_3)$		$a_{13}(\vec{x}_3)$		$a_{23}(x_3)$
wind @ x_4		$W(t, \vec{x}_4)$		$a_{14}(\vec{x}_4)$		$a_{24}(x_4)$
wind @ x_5		$W(t, \vec{x}_5)$		$a_{15}(\vec{x}_5)$		$a_{25}(x_5)$
sealvel @ x_6		$h_e(t, \vec{x}_6)$		$a_{16}(\vec{x}_6)$		$a_{26}(x_6)$
sealvel @ x_7		$\eta(t, \vec{x}_7)$		$a_{17}(\vec{x}_7)$		$a_{27}(x_7)$
rainfall @ x_8		$R(t, \vec{x}_8)$		$a_{18}(\vec{x}_8)$		$a_{28}(x_8)$
⋮		⋮		⋮		⋮

examples of NOAA altimeter data Pacific
 Menemen + Chant (2000)

need to interpret statistical

ex.

" mode 1 explains 50% of the variance"
 " mode 2 explains 30% of variance"
 + higher "modes"

end #19

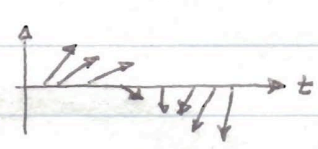
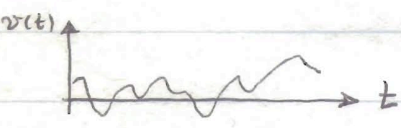
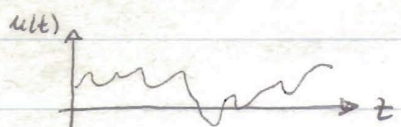
modes physically

→ not always easy



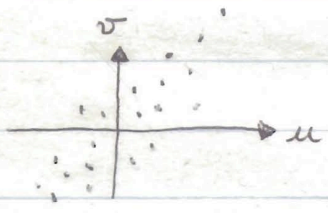
stunt class #20

example: $M=2$ $u_1(t) = u_1(t, x_1) \equiv u(t)$ east component
 $u_2(t) = u_2(t, x_2) \equiv v(t)$ north component
of velocity



North
East

time series of
a vector (velocity)
with $M=2$ components



Let's do an EOF of these and see what we will get

(a) compute the cross-covariance matrix

• remove mean $u' = u - \langle \bar{u} \rangle$
 $v' = v - \langle \bar{v} \rangle$

$$\langle \bar{u} \rangle = \frac{1}{K} \sum_{k=1}^N u(t_k)$$

$$\langle \cdot \rangle = \frac{1}{K} \sum_{k=1}^N \cdot(t_k)$$

$$\begin{pmatrix} \langle u' u' \rangle & \langle u' v' \rangle \\ \langle v' u' \rangle & \langle v' v' \rangle \end{pmatrix} = R_{ij} \begin{matrix} \text{Reynold's} \\ \text{stress tensor} \end{matrix}$$

(b) find eigenvalues of the cross-covariance matrix (Reynold's stress tensor),
i.e.

$$\begin{pmatrix} \langle u' u' \rangle & \langle u' v' \rangle \\ \langle v' u' \rangle & \langle v' v' \rangle \end{pmatrix} \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} = \lambda \begin{pmatrix} e_1 \\ e_2 \end{pmatrix}$$

where λ is an eigenvalue and (e_1, e_2) is an eigenvector

$$\begin{pmatrix} \langle u'u' \rangle - \lambda & \langle u'v' \rangle \\ \langle v'u' \rangle & \langle v'v' \rangle - \lambda \end{pmatrix} \begin{pmatrix} \bar{r}_1 \\ \bar{r}_2 \end{pmatrix} = 0$$

Cramer's rule states that solutions to linear equations are the ratio of 2 determinants, the denominator is the determinant of the matrix and the numerator is the matrix with the i -th column replaced by the vector on the right hand side, i.e.,

$$\begin{pmatrix} 4 & 10 \\ 10 & 30 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 38 \\ 110 \end{pmatrix}$$

$$x_1 = \frac{\det \begin{pmatrix} 38 & 10 \\ 110 & 30 \end{pmatrix}}{\det \begin{pmatrix} 4 & 10 \\ 10 & 30 \end{pmatrix}} = \frac{38 \cdot 30 - 10 \cdot 110}{4 \cdot 30 - 10 \cdot 10} = \frac{40}{20} = 2$$

$$x_2 = \frac{\det \begin{pmatrix} 4 & 38 \\ 10 & 110 \end{pmatrix}}{\det \begin{pmatrix} 4 & 10 \\ 10 & 30 \end{pmatrix}} = \frac{4 \cdot 110 - 10 \cdot 38}{4 \cdot 30 - 10 \cdot 10} = \frac{60}{20} = 3$$

Apply this to our eigenvalue problem

$$e_1 = \frac{\det \begin{pmatrix} 0 & R_{12} \\ 0 & R_{22} \end{pmatrix}}{\det \begin{pmatrix} R_{11} - \lambda & R_{12} \\ R_{21} & R_{22} - \lambda \end{pmatrix}} = \frac{0}{(R_{11} - \lambda)(R_{22} - \lambda) - R_{12} R_{21}}$$

$$e_2 = \frac{\det \begin{pmatrix} R_{11} & 0 \\ R_{21} & 0 \end{pmatrix}}{\det \begin{pmatrix} R_{11} - \lambda & R_{12} \\ R_{21} & R_{22} - \lambda \end{pmatrix}} = \frac{0}{(R_{11} - \lambda)(R_{22} - \lambda) - R_{12} R_{21}}$$

For nontrivial, i.e. $(e_1, e_2) \neq 0$ solutions we need

$$(R_{11} - \lambda)(R_{22} - \lambda) - R_{12}R_{21} = 0$$

$$R_{11}R_{22} - \lambda R_{22} - \lambda R_{11} - \lambda^2 - R_{12}R_{21} = 0$$

correction
Mar. 8, 2002
JL

$$\lambda^2 + (R_{11} + R_{22})\lambda + (R_{12}R_{21} - R_{11}R_{22}) = 0$$

$$\lambda_{1/2} = \frac{-(R_{11} + R_{22}) \pm \sqrt{(R_{11} + R_{22})^2 - 4(R_{12}R_{21} - R_{11}R_{22})}}{2}$$

generally complex, however, it can be shown that for symmetric ($R_{12} = R_{21}$) and Hermitian matrices the eigenvalues are real.

$$(R_{12} = R_{21}^*)$$

Let's take $R_{ij} = \begin{pmatrix} 4 & 10 \\ 10 & 30 \end{pmatrix}$ as before then

$$\lambda_1 = 33.4$$

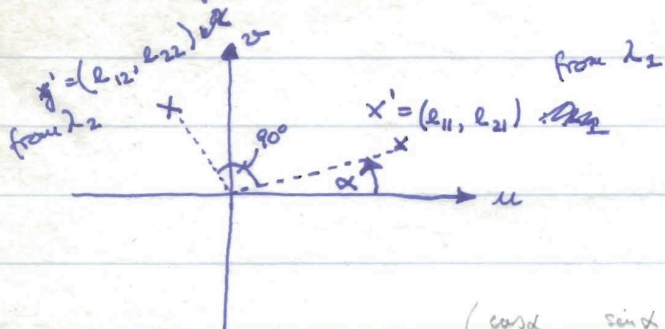
$$\lambda_2 = 0.6$$

note that $\lambda_1 + \lambda_2 = R_{11} + R_{22}$

In order to get the eigenvectors (e_1, e_2) we need to solve for each eigenvalue the linear equations

$$\begin{pmatrix} R_{12} - \lambda_{in} & R_{12} \\ R_{21} & R_{22} - \lambda_{in} \end{pmatrix} \begin{pmatrix} e_{1in} \\ e_{2in} \end{pmatrix} = 0 \quad in = 1, 2$$

What does this now mean, how to interpret the eigenvectors and eigenvalues?



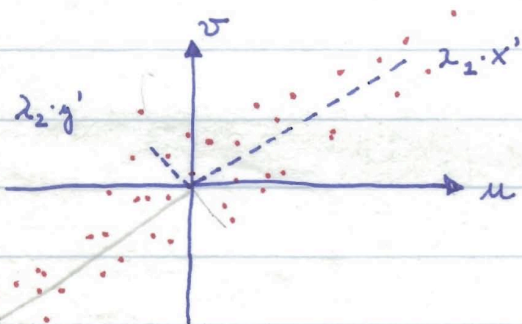
that's a rotation α of the co-ordinate system.

$$\begin{pmatrix} \cos \alpha & \sin \alpha \\ -\sin \alpha & \cos \alpha \end{pmatrix} \begin{pmatrix} u(t) \\ v(t) \end{pmatrix} = \begin{pmatrix} A_1(t) \\ A_2(t) \end{pmatrix} \quad \text{i.e., the eigenfunctions are } \{\sin \alpha, \cos \alpha\}$$

where $A_1(t)$ and $A_2(t)$

are the "orthogonal" the eigenvectors define a "new" base, i.e., they rotate the timeseries of velocities in the (x', y') frame. For symmetric matrices the new base vectors are orthogonal, i.e., they are linear independent, i.e., they are mutually uncorrelated.

the eigenvalues represent the length, magnitude, relative importance of the respective "base" vectors which is closely related to the variance as $R_{11} + R_{22} = \lambda_1 + \lambda_2$



that's an ellipse with a major and a minor axis $\lambda_1 x'$ and $\lambda_2 y'$, respectively

in the "new" system the covariance matrix is $\begin{pmatrix} \sigma & 0 \\ 0 & \lambda_2 \end{pmatrix}$

• data

⇒ principal components of variability (u and v are correlated) orthogonal base representing uncorrelated variability (~~un~~)