



Home Page

Title Page

Contents

◀

▶

◀

▶

Page 27 of 55

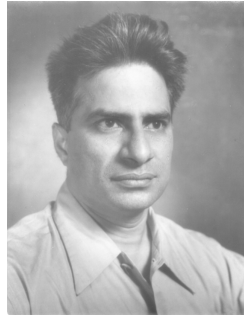
Go Back

Full Screen

Close

Quit

The 'KLT': Introduction



D. D. Kosambi
[1907-1966]
1943

<https://upload.wikimedia.org/wikipedia/commons/0/0f/Kosambi-dd.jpg>

K. Karhunen
[1915-1992]
1945



M. Loève
[1907-1979]
1948

https://upload.wikimedia.org/wikipedia/commons/d/d0/Michel_Lo%C3%A8ve.jpg



H. Hotelling
[1895-1973]

https://upload.wikimedia.org/wikipedia/en/4/49/Harold_Hotelling.jpg

People call it names!

- Karhunen-Loeve Transform
- Hotelling Transform
- Principal Component Analysis
- Eigenvalue-Eigenvector Transform

[Home Page](#)[Title Page](#)[Contents](#)[Page 28 of 55](#)[Go Back](#)[Full Screen](#)[Close](#)[Quit](#)

Pattern Recognition Terms

- A 'pattern' is a $k \times 1$ column vector - a 1-D signal can be represented as a 'pattern'. A $k_1 \times k_2$ 2-D signal (an image) can be represented as a 'pattern' by taking all pixels in raster scan order (row major order) to form a $k \times 1$ 'pattern', $k = k_1 \cdot k_2$
- k -dimensional 'patterns' \mathbf{p}_i^* , $1 \leq i \leq n$
- Stack them up together (in any order) to form a $k \times n$ **Pattern Matrix \mathbf{P}^***



Home Page

Title Page

Contents



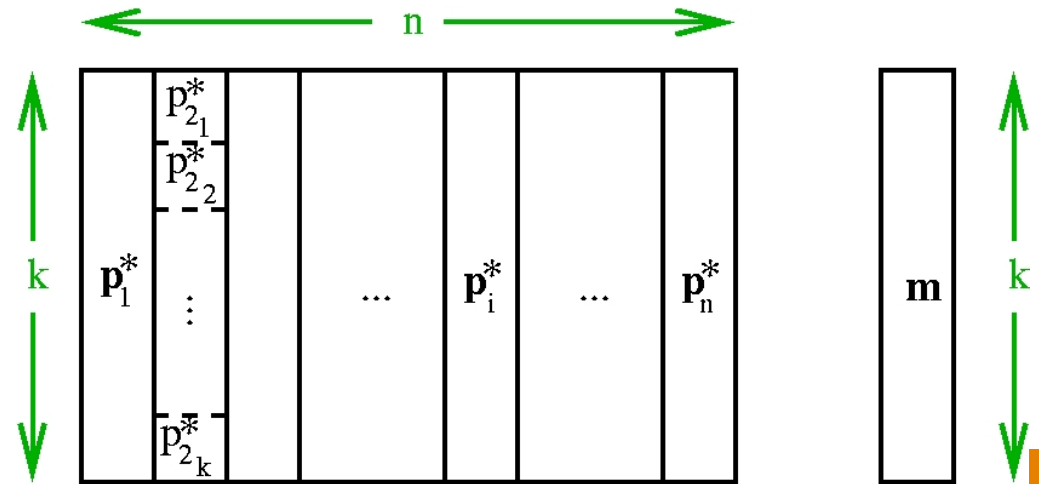
Page 29 of 55

Go Back

Full Screen

Close

Quit



- Normalise each pattern: $\mathbf{p}_i \triangleq \mathbf{p}_i^* - \mathbf{m}$
- $\mathbf{A} \triangleq \frac{1}{n} \mathbf{P} \mathbf{P}^T$: The Covariance Matrix
- Stack together EigenVectors \mathbf{u}_i of \mathbf{A} in decreasing order of the corresponding EigenValues to get the $k \times k$ matrix \mathbf{U}

[Home Page](#)[Title Page](#)[Contents](#)[Page 30 of 55](#)[Go Back](#)[Full Screen](#)[Close](#)[Quit](#)

Linear Algebra Fundamentals

- Phys significance of Eigenvalues & Eigenvectors
- Similar Matrices
- Diagonalisation of a $k \times k$ matrix
- Gram-Schmidt Orthogonalisation
- Eigenvalues of a symmetric real matrix are real
- Eigenvecs of a symmetric matrix: orthonormality

Phys Sig of E'values, E'vectors

- For a $k \times k$ matrix \mathbf{B} , if $\mathbf{B}\mathbf{u}_i = \lambda_i \mathbf{u}_i$, λ_i are the eigenvalues, and \mathbf{u}_i , the corresponding eigenvectors
- Phys sig: matrix \times vector \equiv scaling it!
- Computing eigenvalues: $\mathbf{B}\mathbf{u} - \lambda \mathbf{u} = \mathbf{0} \implies (\mathbf{B} - \lambda \mathbf{I})\mathbf{u} = \mathbf{0} \implies$ non-trivial solution: $|\mathbf{B} - \lambda \mathbf{I}| = 0$
- E'vecs: not unique! Scaled versions also e'vecs

[Home Page](#)[Title Page](#)[Contents](#)[Page 31 of 55](#)[Go Back](#)[Full Screen](#)[Close](#)[Quit](#)

Some Less Important Properties

- $\text{Rank}(\mathbf{B}) = \# \text{ of non-zero eigenvals}$
- $\sum \lambda_i = \text{Trace}(\mathbf{B})$ (sum of main diag), $\prod \lambda_i = |\mathbf{B}|$
- A square matrix \mathbf{A} and \mathbf{A}^T have the same eigenvalues (but usually, different eigenvectors)
 $|\mathbf{A}^T - \lambda \mathbf{I}| = |\mathbf{A}^T - \lambda \mathbf{I}^T| = |(\mathbf{A} - \lambda \mathbf{I})^T| = |\mathbf{A} - \lambda \mathbf{I}|$
- The eigenvalues of a diagonal matrix are those!
eigenvalues: $|\mathbf{B} - \lambda \mathbf{I}| = 0$, $\prod (b_{ii} - \lambda_i) = 0$
- $\mathbf{B}_{k \times k}$ is invertible iff 0 isn't an eigenvalue. eigenvalue 0 iff $|\mathbf{B} - 0\mathbf{I}| = 0$ iff $|\mathbf{B}| = 0$ i.e., non-invertible
- If \mathbf{B} has an eigenvalue-eigenvector pair (λ, \mathbf{u}) , then \mathbf{B}^n ($n \in \mathcal{N}$) has the pair (λ^n, \mathbf{u}) .
 $\mathbf{B}_{k \times k} \mathbf{u}_{k \times 1} = \lambda \mathbf{u}_{k \times 1}$, $\mathbf{B}\mathbf{B}\mathbf{u} = \lambda \mathbf{B}\mathbf{u}$, $\mathbf{B}^2 \mathbf{u} = \lambda^2 \mathbf{u}$, etc
- If \mathbf{B} has an eigenvalue-eigenvector pair (λ, \mathbf{u}) , then \mathbf{B}^{-1} has the pair $(\lambda^{-1}, \mathbf{u})$.
 $\mathbf{B}_{k \times k} \mathbf{u}_{k \times 1} = \lambda \mathbf{u}_{k \times 1}$, $\mathbf{B}^{-1} \mathbf{B} \mathbf{u} = \lambda \mathbf{B}^{-1} \mathbf{u}$, $(1/\lambda) \mathbf{u} = \mathbf{B}^{-1} \mathbf{u}$

[Home Page](#)[Title Page](#)[Contents](#)[◀](#)[▶](#)[◀](#)[▶](#)[Page 32 of 55](#)[Go Back](#)[Full Screen](#)[Close](#)[Quit](#)

- Eigenvectors of a matrix with distinct eigenvalues are linearly independent: Can form a basis

Proof by Contradiction: Suppose not. 'Thin out' this to l indep eigenvectors $\mathbf{u}_1, \dots, \mathbf{u}_l \equiv \lambda_1, \dots, \lambda_l$

Suppose \mathbf{u} was 'thinned out' $\mathbf{u} = \sum_{j=1}^l c_j \mathbf{u}_j$ (1)

1. Multiply (1) by \mathbf{B} : $\mathbf{B}\mathbf{u} = \sum c_j (\mathbf{B}\mathbf{u}_j)$, $\lambda \mathbf{u} = \sum c_j \lambda_j \mathbf{u}_j$

2. Multiply (1) by λ : $\lambda \mathbf{u} = \sum c_j \lambda \mathbf{u}_j$

Subtract: $\mathbf{0} = \sum c_j (\lambda - \lambda_j) \mathbf{u}_j$. Hence, $\forall j$:

$c_j = 0$ (no!) or $\mathbf{u}_j = \mathbf{0}$ (no, as eigenvector is a non-trivial solution) or $\lambda = \lambda_j$ (no!): Contradiction!

- Eigenvalues of a symmetric real matrix are real

$\mathbf{A}\mathbf{u} = \lambda \mathbf{u}$ and $\mathbf{A}^* \mathbf{u}^* = \lambda^* \mathbf{u}^*$, $\mathbf{A}^* = \mathbf{A}$: real

Pre-multiply by \mathbf{u}^{*T} and \mathbf{u}^T , and subtract:

$\mathbf{u}^{*T} \mathbf{A} \mathbf{u} - \mathbf{u}^T \mathbf{A} \mathbf{u}^* = \lambda \mathbf{u}^{*T} \mathbf{u} - \lambda^* \mathbf{u}^T \mathbf{u}^*$ LHS: Consider $(\mathbf{u}^{*T} \mathbf{A} \mathbf{u})^T$, scalar's transpose. $= \mathbf{u}^T \mathbf{A} \mathbf{u}^*$. LHS = 0

RHS: $\mathbf{u}^{*T} \mathbf{u}$: sum-of-sq $\neq 0$ unless all 0, e'vec $\neq \mathbf{0}$

e.g., $[a - jb \ c - jd][a + jb \ c + jd]^T = a^2 + b^2 + c^2 + d^2$

Hence $\lambda = \lambda^*$, only possible if λ is real. QED