## STAT 3690 Lecture 22

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## Sample PCA

- Data  $\mathbf{X} = [\mathbf{X}_1, \dots, \mathbf{X}_n]_{n \times n}^{\top}$ 
  - Each row  $\mathbf{X}_i \stackrel{\mathrm{iid}}{\sim} (\boldsymbol{\mu}, \boldsymbol{\Sigma})$
- Estimate the loadings  $w_i$  through the eigenvectors of sample covariance matrix S or sample correlation

$$\hat{\mathbf{R}} = \begin{bmatrix} \{\widehat{\text{var}}(X_1)\}^{-1/2} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \{\widehat{\text{var}}(X_p)\}^{-1/2} \end{bmatrix} \mathbf{S} \begin{bmatrix} \{\widehat{\text{var}}(X_1)\}^{-1/2} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \{\widehat{\text{var}}(X_p)\}^{-1/2} \end{bmatrix}$$

• Matrix of scores of the first s principal components

$$\mathbf{Z} = [Z_{ij}]_{n \times s} = \widetilde{\mathbf{X}} \widehat{\mathbf{W}}$$

- $-\tilde{\mathbf{X}} = [\mathbf{X}_1 \bar{\mathbf{X}}, \dots, \mathbf{X}_n \bar{\mathbf{X}}]_{n \times p}^{\top}$ : row-centered  $\mathbf{X}$  (i.e. the sample mean has been subtracted from each row of  $\mathbf{X}$ )  $* \bar{\mathbf{X}} = n^{-1} \sum_{i=1}^{n} \mathbf{X}_i$
- $-\widehat{\mathbf{W}} = [\widehat{\boldsymbol{w}}_1, \dots, \widehat{\boldsymbol{w}}_s]_{p \times s} : \widehat{\boldsymbol{w}}_j \text{ is the estimate of } \boldsymbol{w}_j \\ -Z_{ij} = (\mathbf{X}_i \bar{\mathbf{X}})^{\top} \widehat{\boldsymbol{w}}_j : \text{ the } j \text{th PC score for the } i \text{th observation}$

## Geometric interpretation of (sample) PCA

- The definition of PCA as a linear combination that maximises variance is due to H. Hotelling (1933, Journal of Educational Psychology, 24, 417–441).
- PCA was introduced earlier by K. Pearson (1901, Philosophical Magazine, Series 6, 2(11), 559–572) to minimize the overall error in reconstructing data points

$$(\bar{\mathbf{X}}, \widehat{\mathbf{W}}, \mathbf{Z}_i) = \arg\min_{\boldsymbol{\theta}, \mathbf{A}, \mathbf{B}_i} \sum_{i=1}^n \|\mathbf{X}_i - \boldsymbol{\theta} - \mathbf{A}\mathbf{B}_i\|^2$$

-  $\mathbf{Z}_i$ : the *i*th row of score matrix  $\mathbf{Z}$