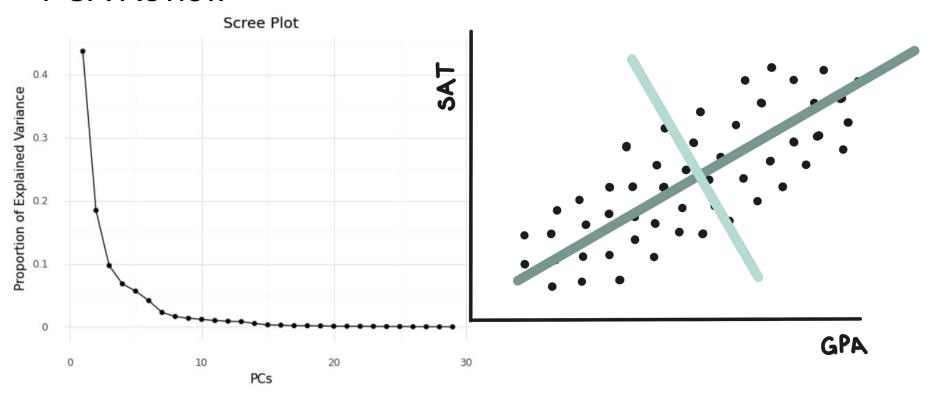
Autoencoders

Dr. Chelsea Parlett-Pelleriti

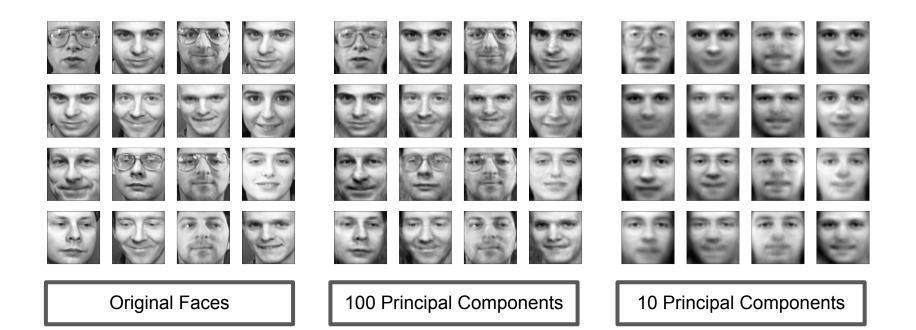
Outline

- PCA
- Architecture
- Nonlinear Encoders/Decoders
- Regularization: Penalization
- Regularization: Penalizing Derivatives
- Regularization: Denoising
- Convolutional Autoencoders
- Example: Image Compression

PCA Review



PCA and Lossy Compression



PCA and Linearity

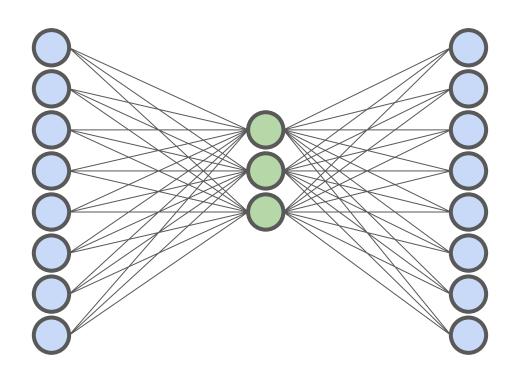
PC's are **linear combinations** of the original variables

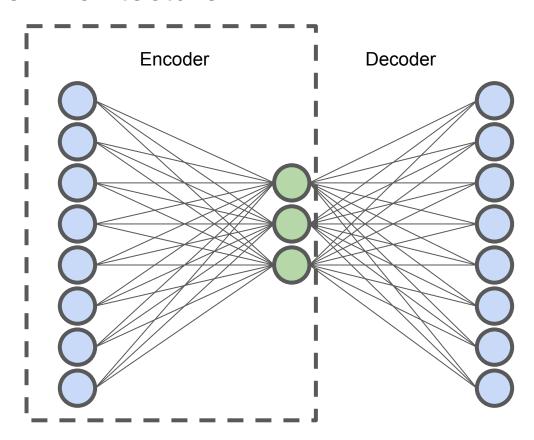
 $w^T x$

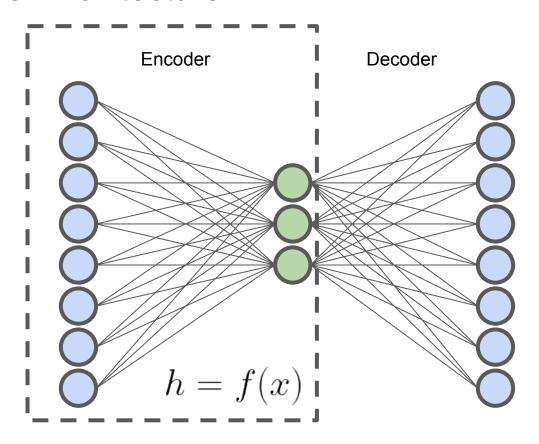
Original Data Eigenvectors/ **Component Scores** Loadings (New Variables) Age Weight Height 🍍 Rate **3** The color of the circles represents data from different people PC₁ Score for $\frac{2}{3}$ The color of the rectangles represents data from different original variables The linetype represents different eigenvectors (the eigenvectors are the loadings for each Principal Component). PC₂ Score for $\frac{1}{100}$ 3 The color of the squares represent which original variable that value goes with. **PCA Matrix Algebra** @chelseaparlett

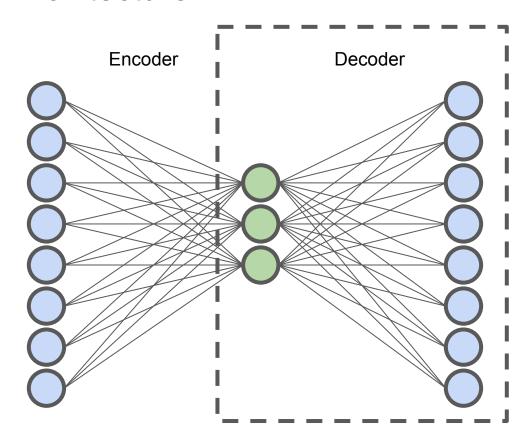
What if we had NON-Linear PCA?

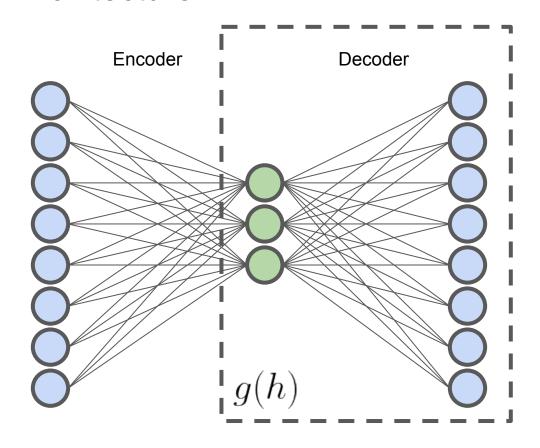


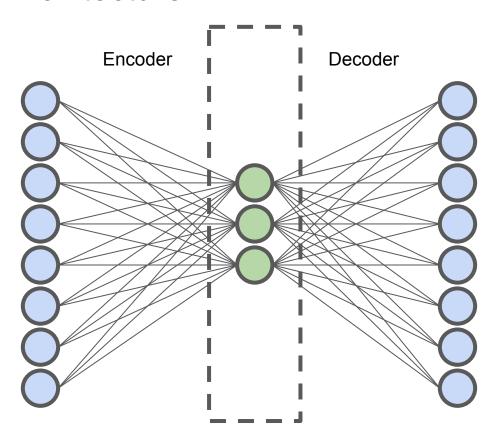


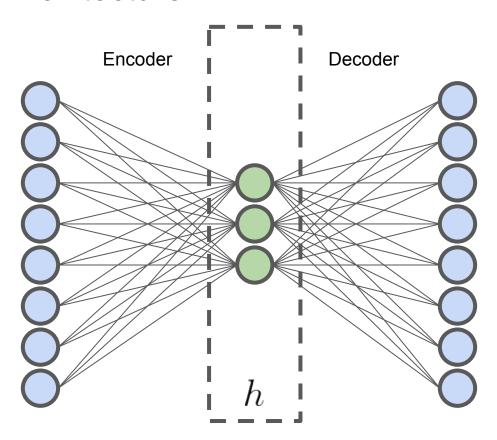


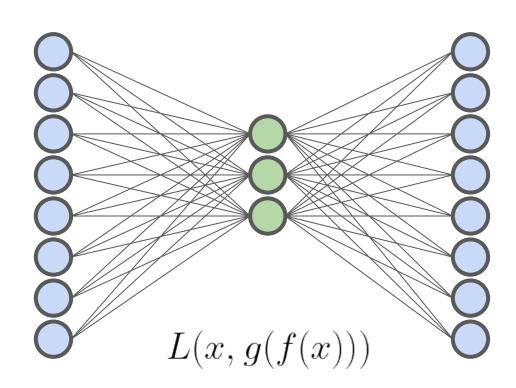








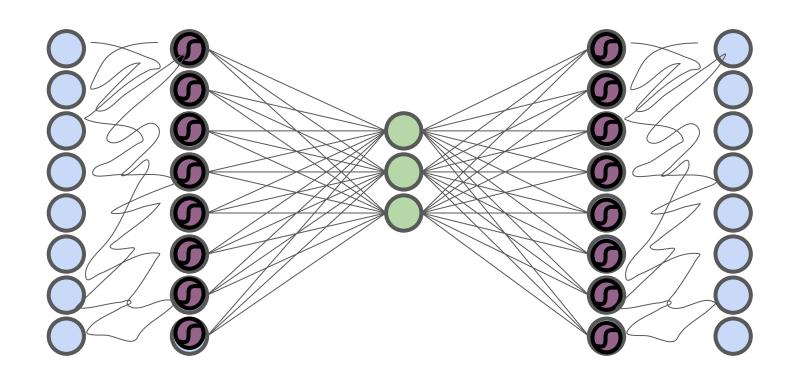




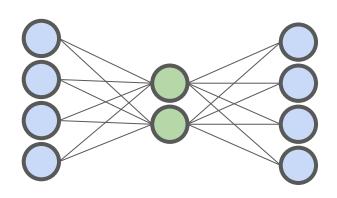
PCA and Autoencoders

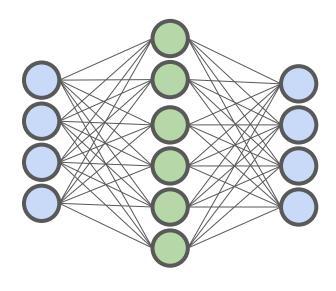
- Encoder is linear
- Loss is MSE

Adding Nonlinearity

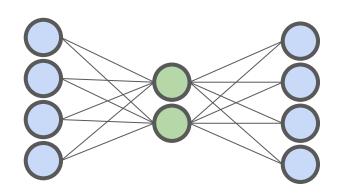


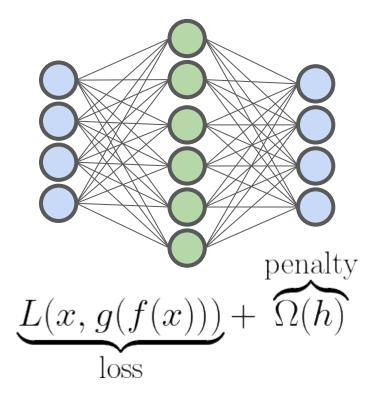
Undercomplete and Overcomplete





Sparse Autoencoders



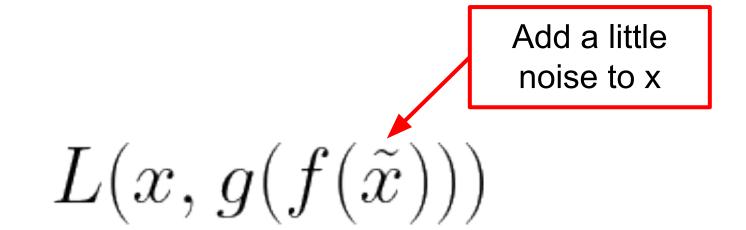


Penalizing Derivatives

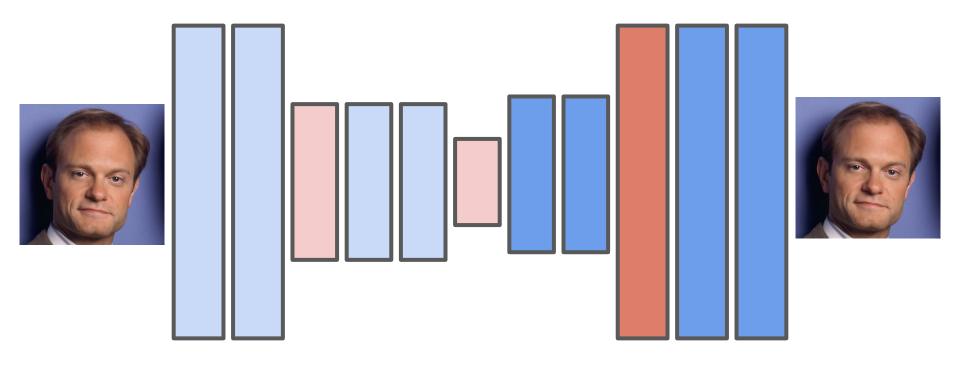
$$L(x, g(f(x))) + \Omega(h, x)$$

$$\Omega(h, x) = \lambda \sum_{i} \|\nabla_x h_i\|^2$$

Denoising Autoencoder (DAE)



Convolutional Autoencoders



Denoising Convolutional Autoencoder

