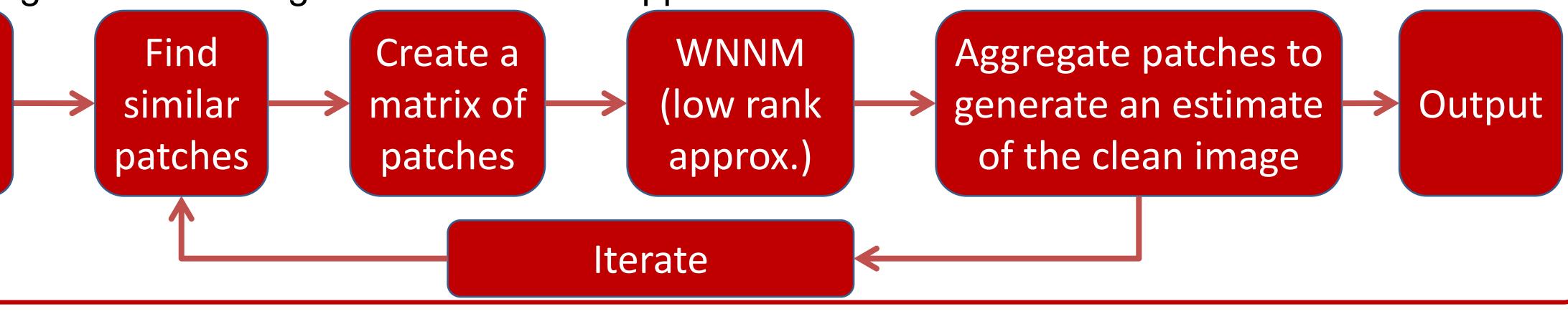
# EE367 Project: Implementation of the weighted nuclear norm minimization for image denoising Andrei Kanavalau

<ul> <li>Motivation</li> <li>Image denoising is an important part of the image processing pipeline</li> <li>Foundational to other tools</li> <li>Gain insight by implementing a state of the art technique based on weighted nuclear norm minimization (WNNM)</li> </ul>	<text></text>
<ul> <li>Related Work</li> <li>Numerous techniques available from simple Gaussian and median filters to denoising convolutional neural networks (DnCNN)</li> <li>Non-local approaches have superior performance [1], especially Block matching and 3D filtering (BM3D) [2], WNNM [3], and DnCNN</li> </ul>	• Com
<b>References</b> [1] L. Fan, F. Zhang, H. Fan, and C. Zhang. Brief review of image denoising techniques. Visual Computing for Industry, Biomedicine, and Art, 2(1):7, 2019. [2] K. Dabov, A. Foi, V. Katkovnik and K. Egiazarian, "Image Denoising by Sparse 3-D Transform-Domain Collaborative Filtering," in IEEE Transactions on Image Processing, vol. 16, no. 8, pp. 2080-2095, Aug. 2007, doi: 10.1109/TIP.2007.901238. [3] S. Gu, L. Zhang, W. Zuo, and X. Feng. Weighted nuclear norm minimization with application to image denoising. In Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR), June 2014.	• Avera PSNR SSIM

Department of Electrical Engineering, Stanford University

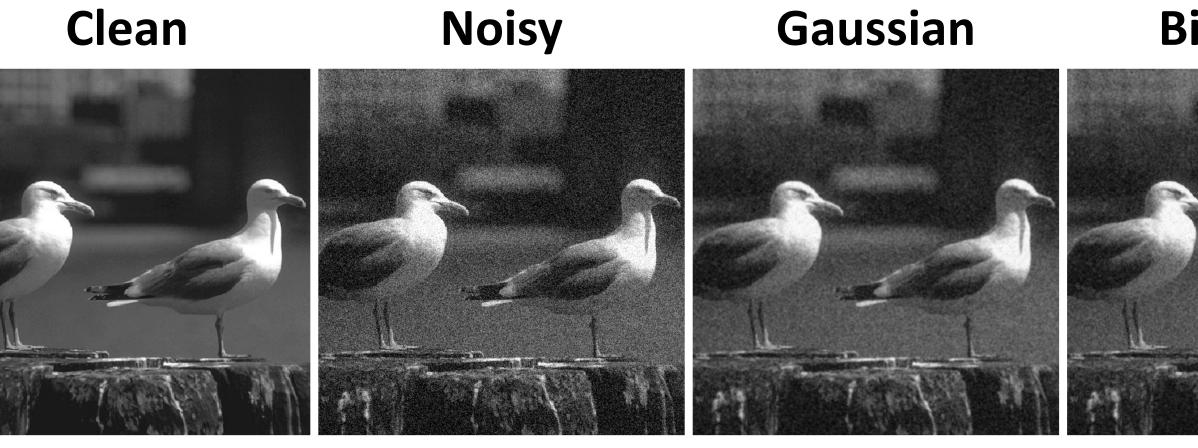
## Method

NM denoising relies of the assumption that the structure in the image can be identified ong the noise using a matrix low rank approximation



### **Experimental Results**

npare to other denoising techniques



20.30 0.39

28.70 0.83

rages across 15 grayscale images:

-	20.22	26.36
-	0.49	0.80

### NLM Bilateral WNNM 31.72 28.41 29.49 0.81 0.88 0.93 26.60 26.97 28.40 0.86 0.80 0.82