



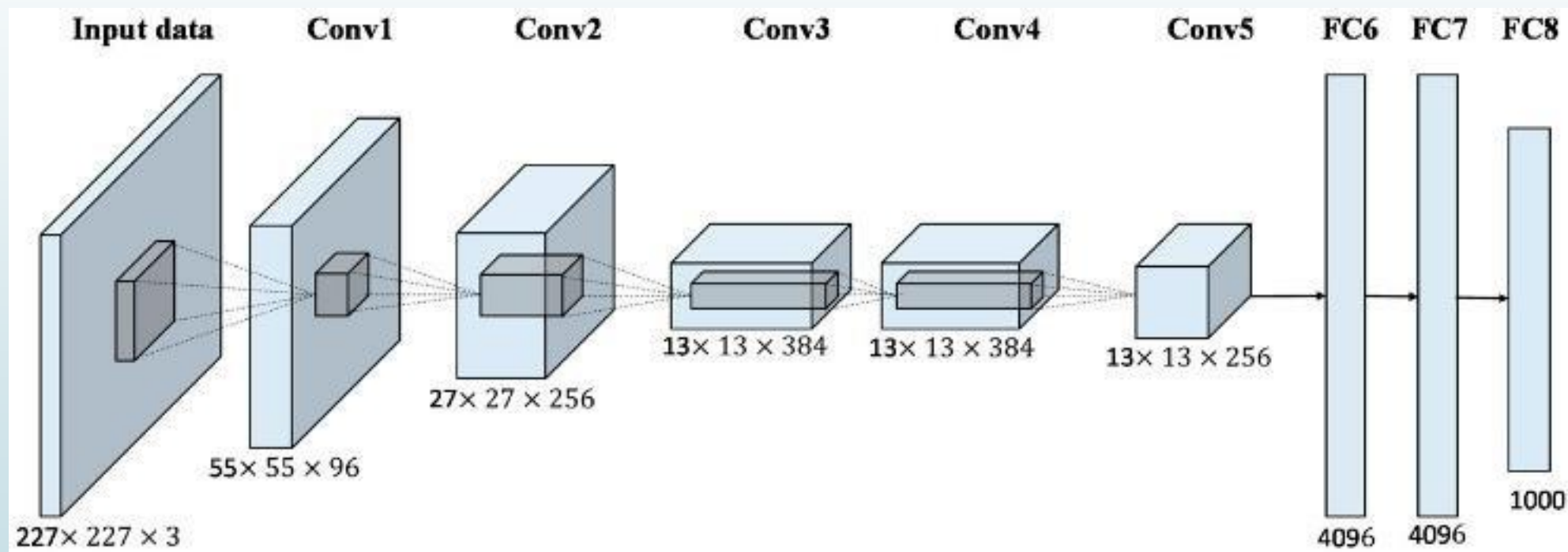
CNN's Family



AlexNet (2012)

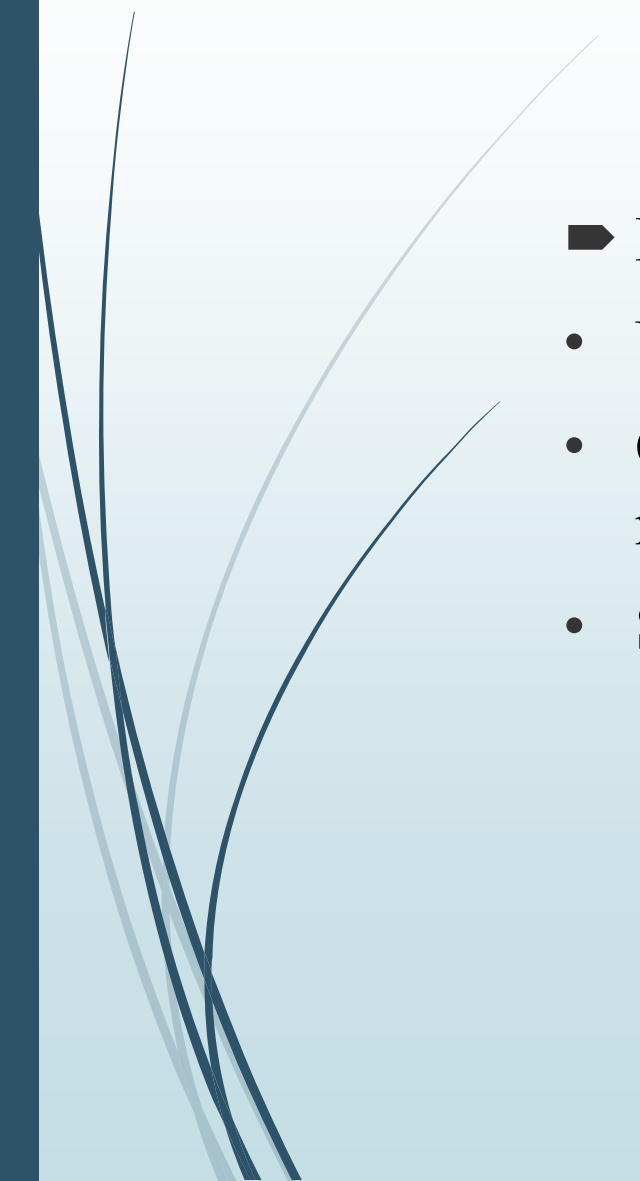
- Won the ImageNet competition in 2012, popularizing deep learning.
- Five convolutional layers followed by three fully connected layers, introduced **ReLU activation** for faster training.
- Used **dropout** for regularization.
- Trained on GPUs for the first time.

AlexNet Architecture

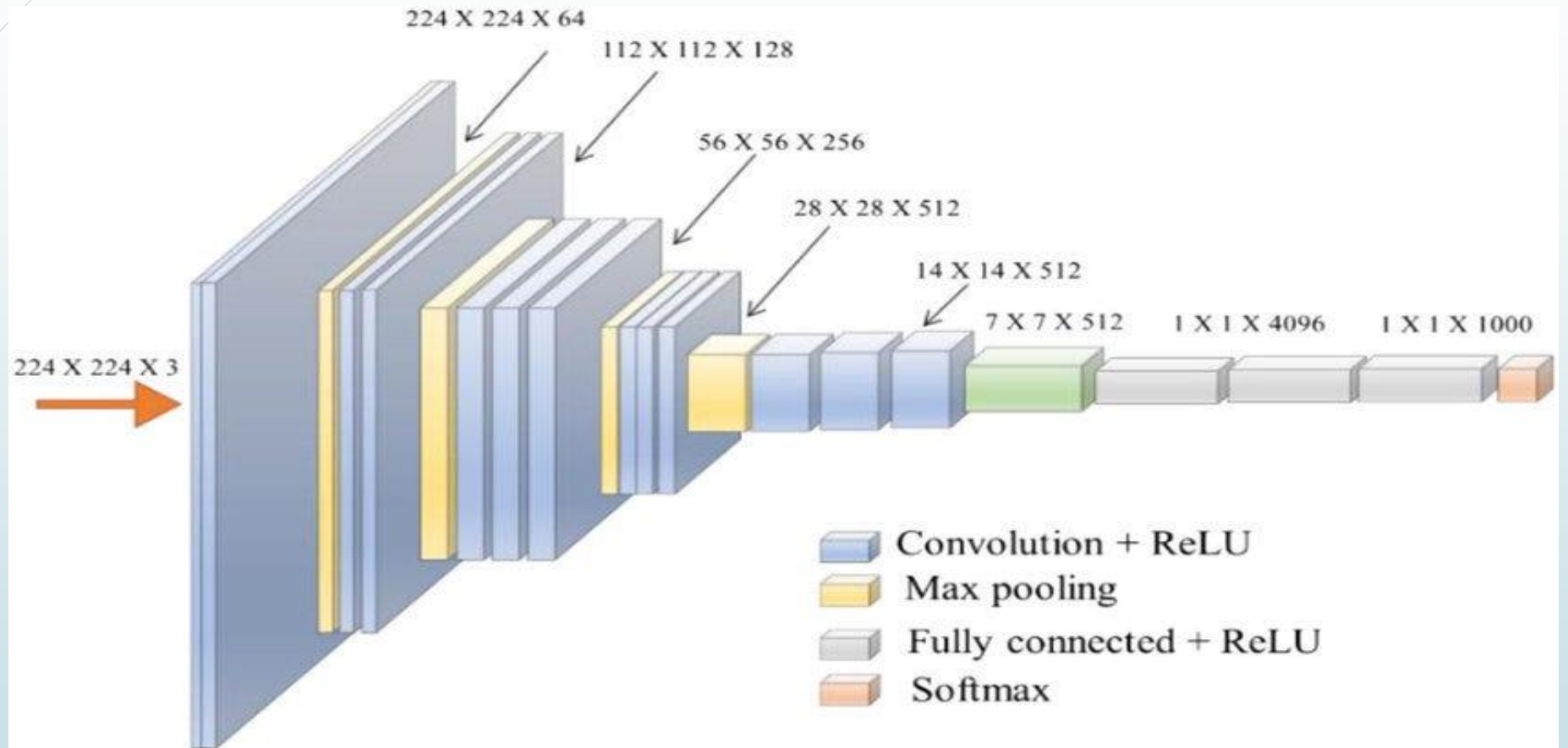




VGGNet (2014)

- Developed by the Visual Geometry Group at Oxford.
 - Used small 3x3 filters throughout the network.
 - Came in two variants: VGG-16 and VGG-19 (16 and 19 layers, respectively).
 - Showed that depth improves performance.
- 

VGGNet Architecture



<https://ai.plainenglish.io/vggnet-with-tensorflow-transfer-learning-with-vgg16-included-7e5f6fa9479a>



ResNet (2015)

- Introduced **residual connections** to enable training of very deep networks
- Used **skip connections** to bypass layers, addressing the vanishing gradient problem.
- Came in variants like ResNet-50, ResNet-101, and ResNet-152.
- Won the ImageNet competition in 2015 with a top-5 error rate of 3.57%



Advanced Architectures


- 
- 1. MobileNet**
 - 2. ShuffleNet**
 - 3. NASNet**
 - 4. Transformer-based Vision Models.**



Data Augmentation

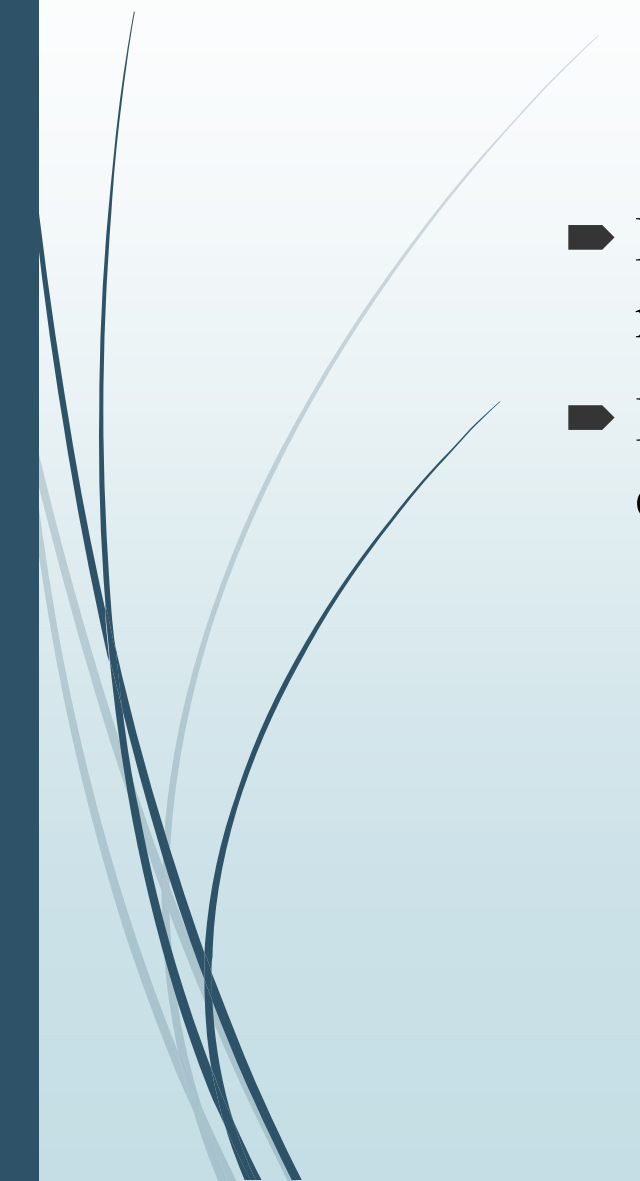


What is Data Augmentation

- Data augmentation is a technique used to artificially increase the size and diversity of a training dataset by applying various transformations to the existing data.
 - It helps improve the generalization of machine learning models, especially in scenarios where labeled data is scarce.
- 



Why is it important

- Prevents overfitting: The model learns to generalize better than memorizing the training data by introducing variations.
 - Improves model robustness: The model becomes more invariant to changes in input data (e.g., rotation, scaling, noise).
- 



Data Augmentation Techniques

- 
- 1. Geometric Transformations**
 - 2. Color Space Transformations**
 - 3. Noise Injection**

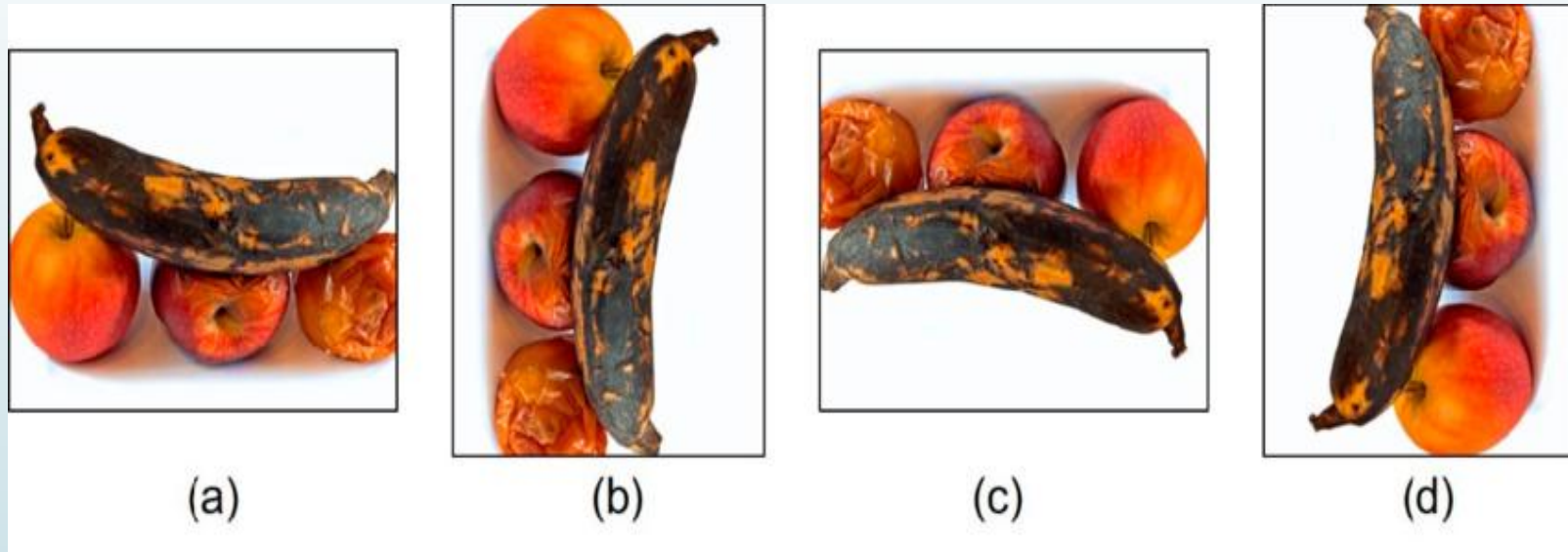


Geometric Transformations



- **Rotation:** Rotate the image by a certain angle (e.g., 90° , 180°)
- **Translation:** Shift the image horizontally or vertically.
- **Scaling:** Zoom in or out of the image.
- **Flipping:** Flip the image horizontally or vertically.
- **Cropping:** Randomly crop a portion of the image.

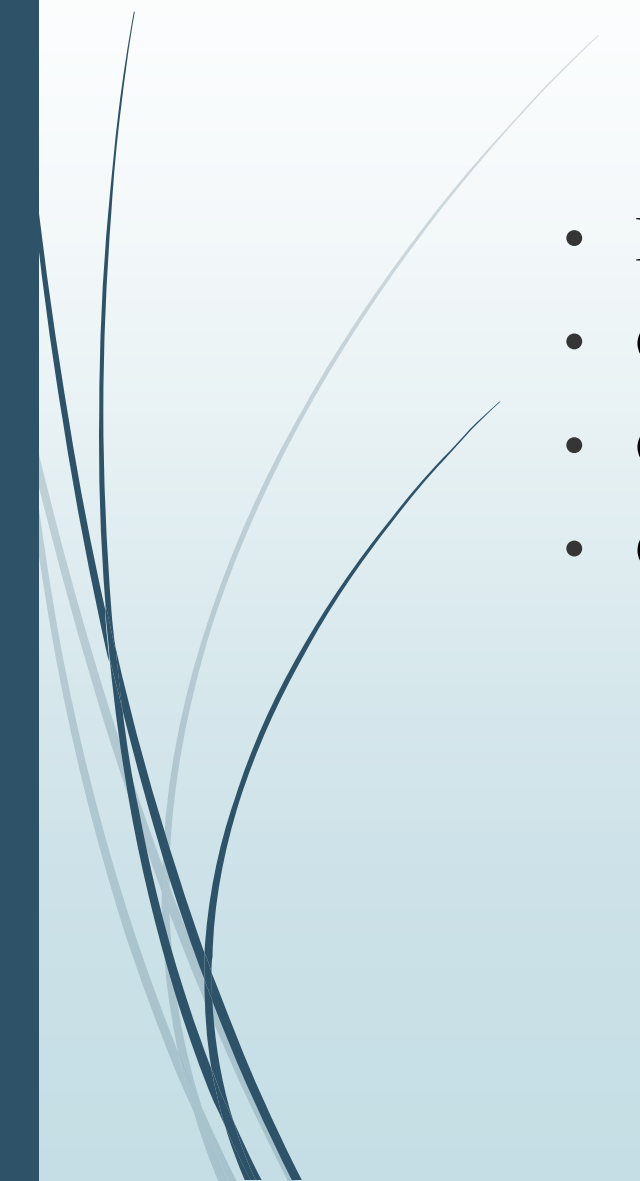
Geometric Transformations



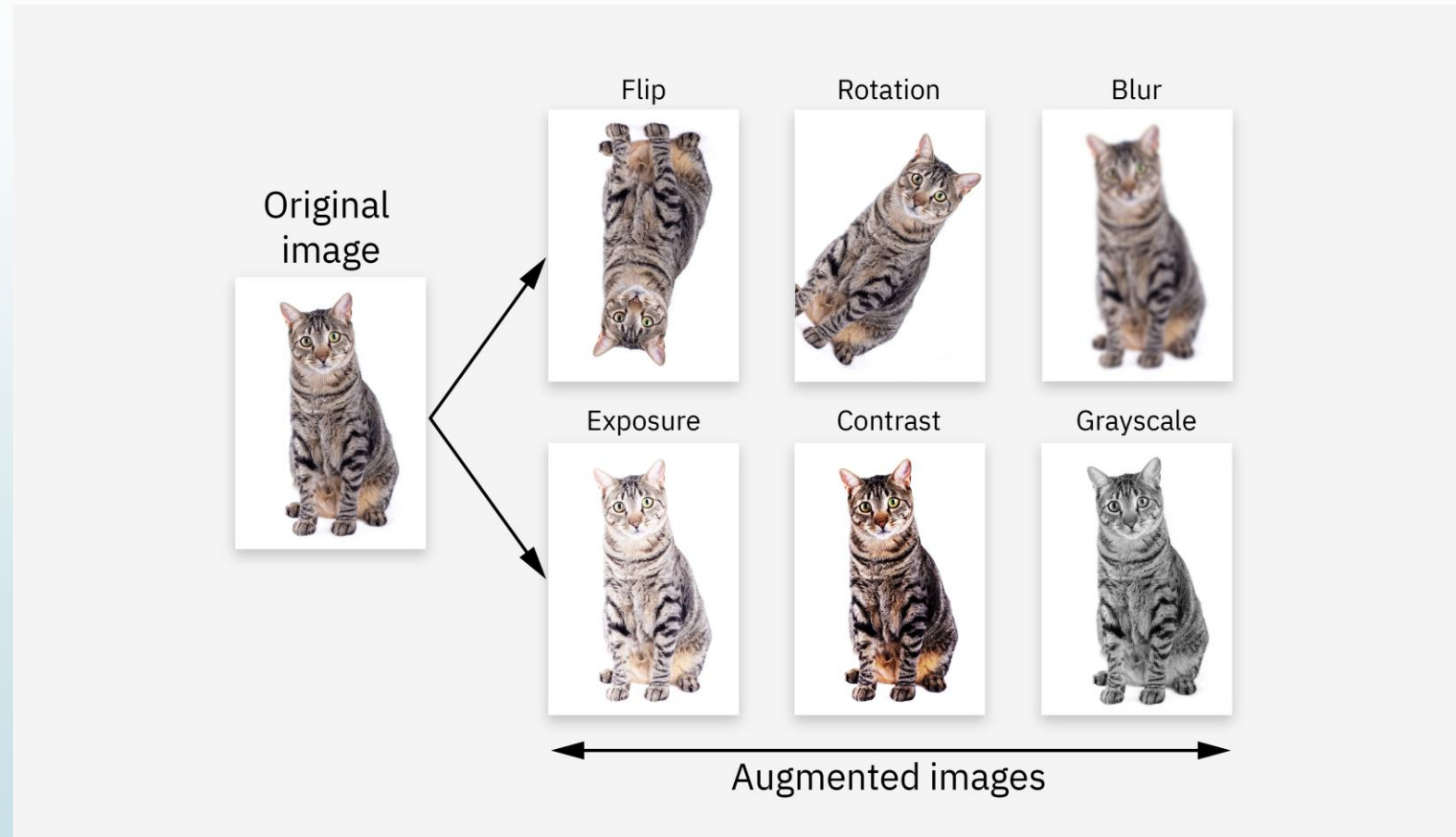
https://www.researchgate.net/publication/364764514_Improved_Classification_Approach_for_Fruits_and_Vegetables_Freshness_Based_on_Deep_Learning/figures?lo=1&utm_source=google&utm_medium=organic



Color Space Transformations

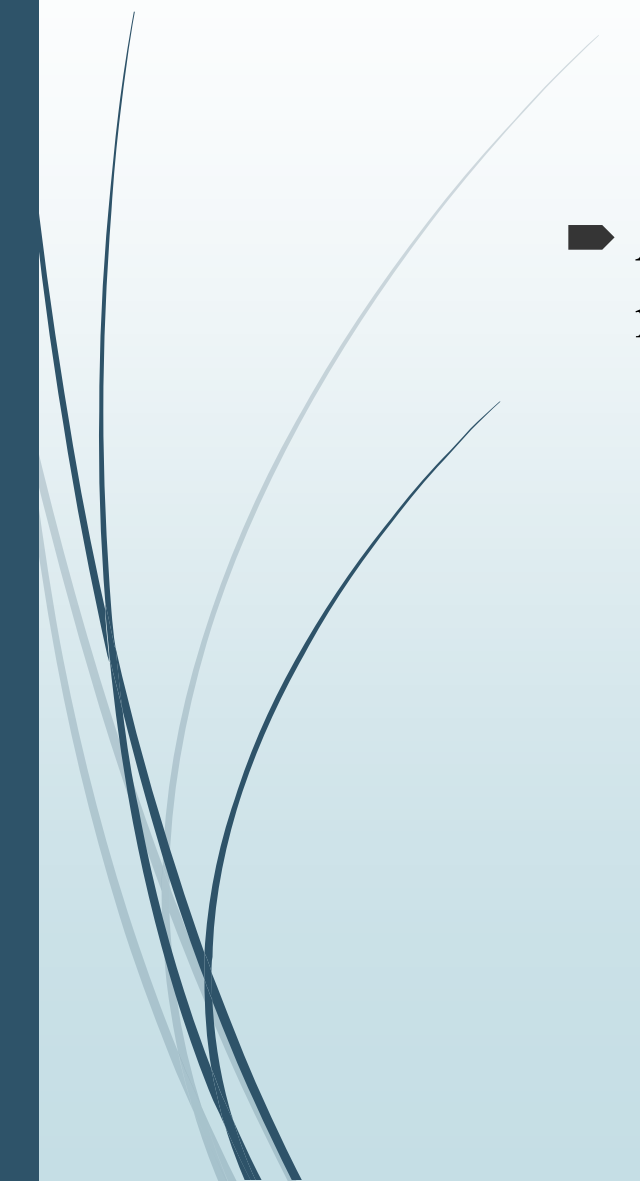
- **Brightness Adjustment:** Increases or decreases the brightness.
 - **Contrast Adjustment:** Modify the contrast of the image.
 - **Color Jittering:** Randomly change the color balance.
 - **Grayscale Conversion:** Convert the image to grayscale.
- 

Color Space Transformations

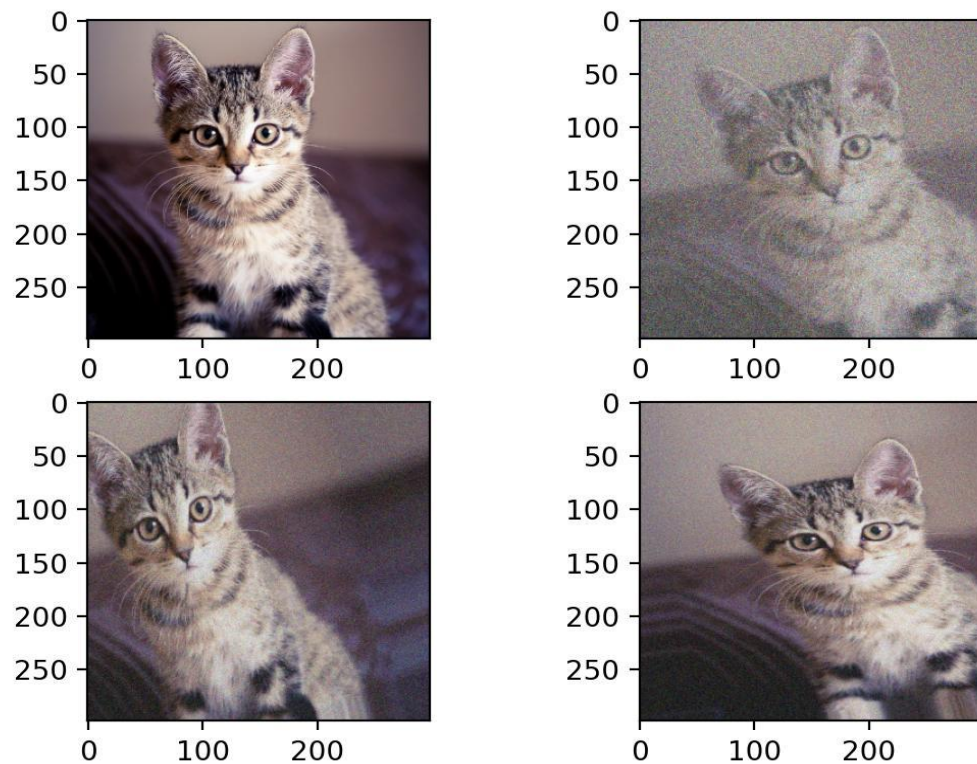




Noise Injection

- Add random noise (e.g., Gaussian noise) to the image to make the model robust to imperfections.
- 

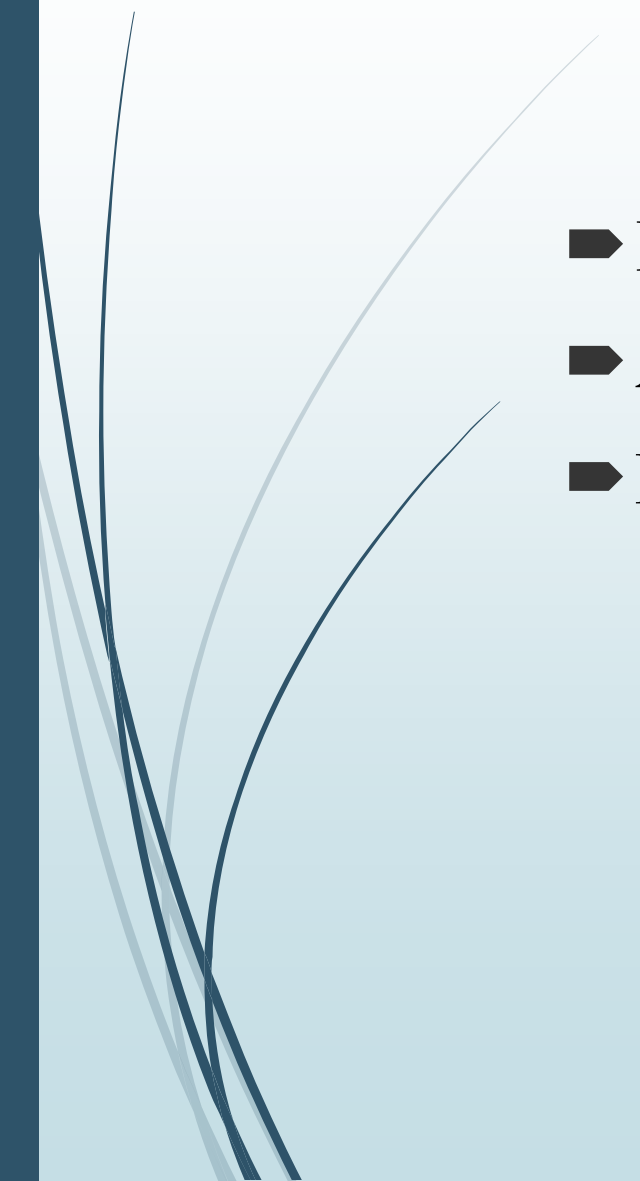
Noise Injection



<https://stackoverflow.com/questions/43382045/keras-realtime-augmentation-adding-noise-and-contrast>

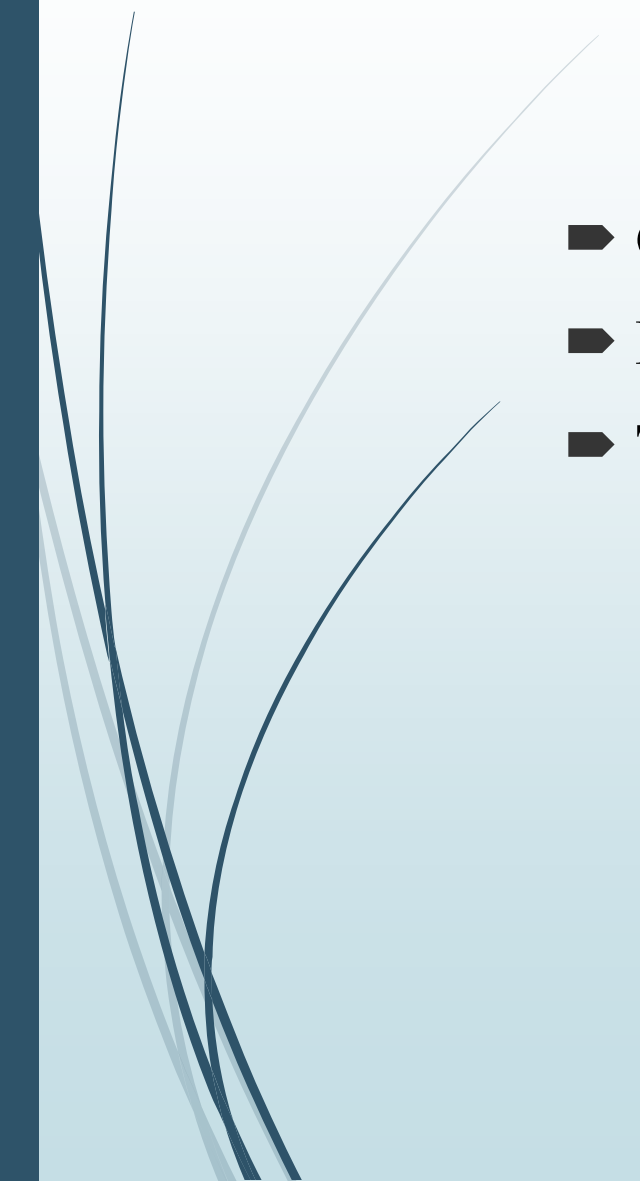


Best Practices for Data Augmentation

- **Preserve Label Integrity**
 - **Avoid Over-augmentation**
 - **Domain-Specific Augmentation**
- 



Challenges and Limitations

- **Computational Cost**
 - **Loss of Information**
 - **Task-Specific Constraints**
- 



Applications of Data Augmentation

- **Computer Vision:** Image classification, object detection, segmentation.
 - **Natural Language Processing:** Text classification, machine translation.
 - **Speech Recognition:** Improving robustness to noise and accents.
 - **Healthcare:** Augmenting medical images for better diagnosis.
- 