Topics in Module-3

- Convolution layers
- Max pool layers,
- ELU Gradient Descent,
- Training CNN-initialization,
- CNN architectures VGG, Google Net,
- ResNet,
- Data augmentation,
- Transfer learning,

Data Augmentation

Augmented vs. Synthetic data

- Augmented data is driven from original data with some minor changes.
 In the case of image augmentation, we make geometric and color space transformations (flipping, resizing, cropping, brightness, contrast) to increase the size and diversity of the training set.
- Synthetic data is generated artificially without using the original dataset. It often uses DNNs (Deep Neural Networks) and GANs (Generative Adversarial Networks) to generate synthetic data.

When to use data augmentation?

- •To prevent models from overfitting.
- •The initial training set is too small.
- •To improve the model accuracy.
- •To reduce the **operational cost of labeling** and cleaning the raw dataset.

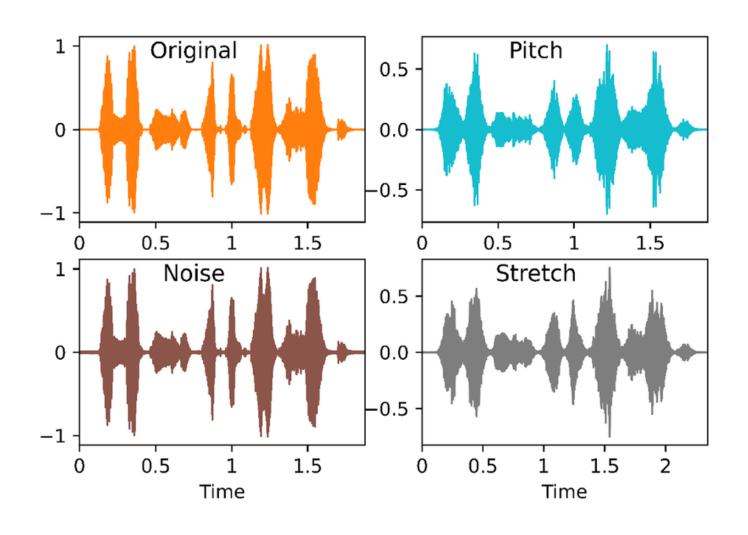
Audio and text Data Augmentation

- Noise injection: add gaussian or random noise to the audio dataset to improve the model performance.
- Shifting: shift audio left (fast forward) or right with random seconds.
- Changing the speed: stretches times series by a fixed rate.
- Changing the pitch: randomly change the pitch of the audio.

Text Data Augmentation

- Word or sentence shuffling: randomly changing the position of a word or sentence.
- Word replacement: replace words with synonyms.
- Syntax-tree manipulation: paraphrase the sentence using the same word.
- Random word insertion: inserts words at random.
- Random word deletion: deletes words at random.

Audio based data augmentation



Text based data augmentation

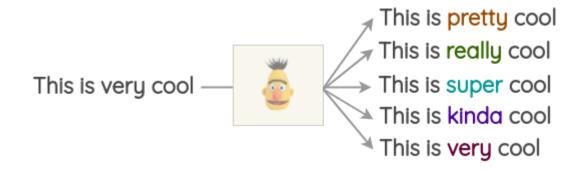
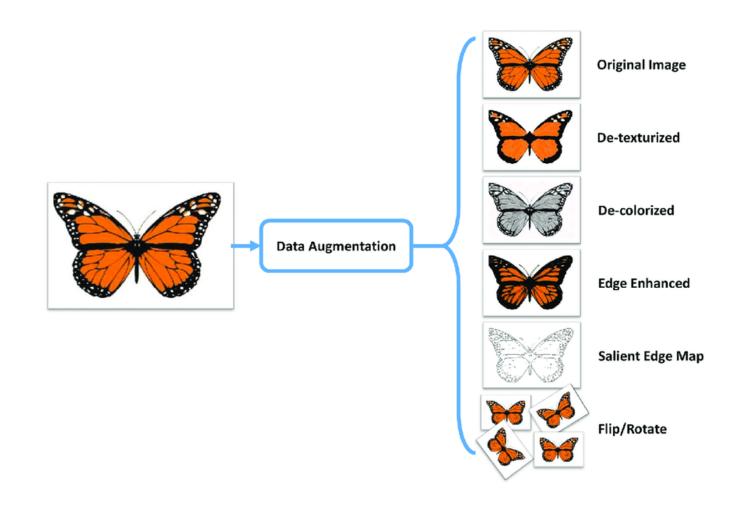


Image Augmentation

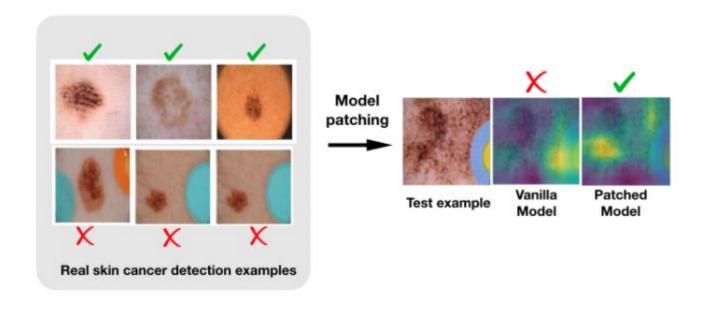
- Geometric transformations: Randomly flip, crop, rotate, stretch, and zoom images. You need to be careful about applying multiple transformations on the same images, as this can reduce model performance.
- Color space transformations: Randomly change RGB color channels, contrast, and brightness.
- Kernel filters: Randomly change the sharpness or blurring of the image.
- Random erasing: Delete some part of the initial image.
- Mixing images: Blending and mixing multiple images.

Image based data augmentation

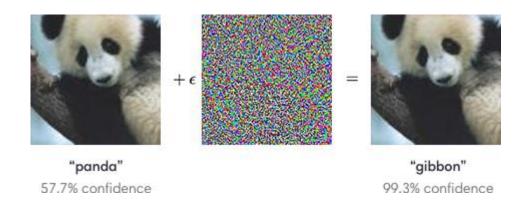


Data augmentation for model patching

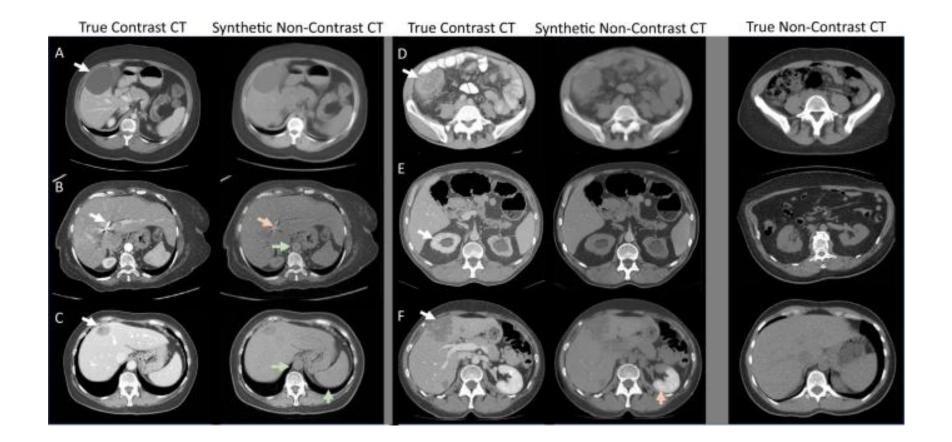
• Model patching is becoming a late-breaking area that would alleviate the major problem in safety-critical systems, including healthcare (e.g. improving models to produce MRI scans free of artifact) and autonomous driving (e.g. improving perception models that may have poor performance on irregular objects or road conditions).



 by adding a small amount of noise to an image can confuse the Al classifier and classifies a panda as a gibbon.



- CT scan images generated by a cycleGAN, which is a variation of GAN.
- This is how GAN-generated CT scan images are being used in the medical field to increase the dataset.
- Once the dataset is created, it can be used for classification or any other task.



Neural Style Transfer-based augmentation

- A series of convolutional layers are trained such that the images are deconstructed where content and style can be separated.
- After separation, the content from an image is composed with the style of another image to create an augmented style image.
- Thus, the content remains the same but the style is changed.
- This increases the robustness of the model as the model is working independently of the style of the image.

