

Additional Case Study

Image Denoising using Autoencoders

Domain

Computer Vision, Image quality

Business Context

Optical Character Recognition (OCR) is the process of getting typed or handwritten documents into a digitised format. If you've read a classic novel on a digital reading device or had your doctor pull up old healthcare records via the hospital computer system, you've probably benefited from OCR.

OCR makes previously static content editable, searchable, and much easier to share. But, a lot of documents eager for digitisation are being held back. Coffee stains, faded sun spots, dog-eared pages, and a lot of wrinkles are keeping some printed documents offline and in the past.

Problem Statement

Getting useful information out of the images which contain a lot of noise is a challenge. In this case study, we will learn how we can use Autoencoders to solve this problem.

Dataset

A dataset of images of scanned text. These images contain various styles of text, to which synthetic noise has been added to simulate real-world, messy artefacts. The training set includes the test without the noise (train_cleaned)

- Noisy images: 144
- Cleaned images: 144
- Test images(noisy): 72

Commented [1]: Add Problem statement.

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Sample Images

Noisy - Train

There exist several methods to design forms to be filled in. For instance, fields may be surrounded by boxes, by light rectangles or by guiding rules to specify where to write and, therefore, to avoid skew and overlapping with other parts of the form. Guides can be located on a separate sheet located below the form or they can be printed on the form. The use of guides on a separate sheet from the point of view of the quality of the output but requires giving more instructions and, therefore, restricts its use to tasks where this type of a

Cleaned - Train

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Noisy - Test

A new offline handwritten database for the Spanish language (Spanish Restricted-domain Task of Cursive Script), has recently been developed: the Spartacus database (Spanish Restricted-domain Task of Cursive Script). There were two main goals in this corpus. First of all, most databases do not contain Spanish sentences. Spanish is a widespread major language. Another important reason is that most databases are not designed for semantic-restricted tasks. These tasks are commonly used in the evaluation of linguistic knowledge beyond the lexicon level in the recognition of handwriting. As the Spartacus database consisted mainly of short sentences and paragraphs, the writers were asked to copy a set of sentences in free line fields in the forms. Next figure shows one of the forms used in the database. These forms also contain a brief set of instructions given to the

Steps

1. Extract data from zip files

2. Look at the dataset in more detail
 - a. Set data directory path variables
 - b. Check the number of images in train, train_cleaned & test folders
 - c. View feature and label
3. Define a function to load the images and save them into NumPy array
 - a. Get the NumPy arrays for features and labels using the above function
4. Split data into training and validation
5. Define the autoencoder model
6. Compile the model
7. Summarize the model
8. Fit the model
9. Predict on test data

Explore more

- Explore if you can implement autoencoders on a different dataset
- Build more complex and deep architectures to get better results

Happy Learning!