



# SYDA model checklist[\[1199\]](#)

| Table no. | Index                  |
|-----------|------------------------|
| 1         | General                |
| 2         | Training Data          |
| 3         | Evaluation Data        |
| 4         | Project config details |
| 5         | Metrics                |
| 6         | Use                    |
| 7         | Clinical Evaluation    |
| 8         | Ethical considerations |
| 9         | Informed consent       |
| 10        | Legal                  |

## 1 General[\[1200\]](#)

|                             |                                                                                       |
|-----------------------------|---------------------------------------------------------------------------------------|
| Project Name                | Synthetic Data Shadow                                                                 |
| Report version              | Version number (can it be maintained somewhere else)                                  |
| Person developing the model | Srinivasan Duraimani                                                                  |
| Date                        | (The date of submitting this DOC)                                                     |
| Model name                  | CT image generator (or could be filename "model.pt or model.h5")                      |
| Model description           | This Model Generates realistische synthetic ct scans given a set of image properties. |
| Model type                  | Data Generator                                                                        |
| Model version               | 0.0                                                                                   |
| Link to Github Repository   | git link to be shared after model artifacts are stored with version control           |

## 2 Training Data[\[1201\]](#)

### 2.1 Data version 0[\[1390\]](#)

Details about Data version 0 can be found [here](#) switch to preview mode to view the full contents

### 2.2 Annotation protocol[\[1204\]](#)

No Annotation was required for this Model. Since the model is trained to only generate realistic images from random Noise inputs.

### 3 Evaluation Data[\[1205\]](#)

Evaluation Data would Not be Applicable for a Generative Model.

#### 3.1 Dataset[\[1206\]](#)

N.A

#### 3.2 Demographic Details and Stats[\[1207\]](#)

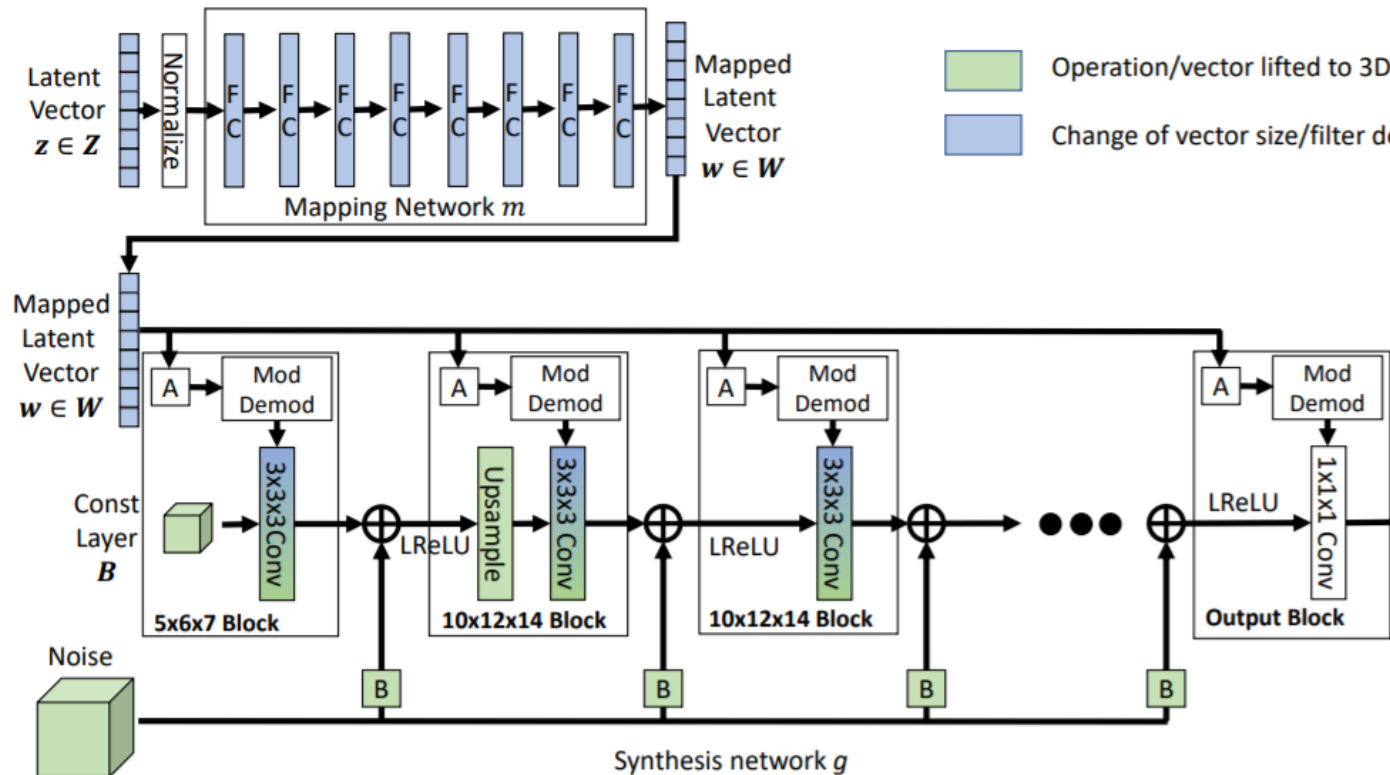
N.A

#### 3.3 Annotation protocol[\[1208\]](#)

N.A

## 4 Project Config[\[1209\]](#)

### 4.1 Algorithm[\[1210\]](#)



The Generative model has 2 models a mapping Network that consists of 8 Linear Layers and a Synthesis Network. The Mapping network Starts with an input of Latent vector  $Z$  in our case its 96, and generates  $W$  which will be 96. The Synthesis Networks Starts from a const  $4 \times 4 \times 4 \times 512$  tensor that is fed into a  $3 \times 3 \times 3$  Convolution layer this similar block is repeated 5 times with Upsampling done before each block. the  $W$  is also injected into the convolution layers. Additional Noise  $B$  is added after the Convolution and before Upsampling. this model generates images of shape  $128 \times 128 \times 128$  pixels with  $1.5 \times 1.5 \times 1.5$  voxel spacing in the range of -1 to 1. which can be converted to a range of -1000 to +1000 to simulate HU values of CT scans.

The Discriminator is a simple convolutional model with downsampling done that predicts the validity of an image as real or fake. The loss is a min max game between the Generator and Discriminator along with path regularization of real images.

### 4.2 Pre-Processing Steps[\[1211\]](#)

We do the following preprocessing steps:

During Training:

- convert the dicom images into niftii's
- resample the niftii images to a consistent 1.5x1.5x1.5 mm voxel spacing
- extract the image data, affine data and image attributes and convert it into a tf record
- tf records as they get extracated in the training code they are checked for voxel spacing of 1.5x1.5x1.5 mm
- the values are scaled from -1000 to +1000, to range -1 to +1, clipping outliers
- the image is resized with a maintained aspect ratio and by th elongest side, to 128
- all sides are then padded with padding value of -1
- converted to tensors for training.

During inference:

- convert the dicom images into niftii's
- resample the niftii images to a consistent 1.5x1.5x1.5 mm voxel spacing
- load the niftii image
- the values are scaled from -1000 to +1000, to range -1 to +1, clipping outliers
- the image is resized with a maintained aspect ratio and by th elongest side, to 128
- all sides are then padded with padding value of -1
- converted to tensors for inference.
- Once the tensors pass through the model creating a generated image of range -1 to +1.

#### 4.3 Libraries Used[\[1212\]](#)

1. torch==1.7.1+cu110
2. torchvision==0.8.2+cu110
3. torchaudio==0.7.2
4. torchmetrics
5. pytorch\_lightning
6. monai[all]
7. addict
8. kornia
9. pycm
10. cloudml-hypertune
11. google-cloud-storage
12. gcsfs
13. tensorflow
14. fsspec[gcs]
15. gsutil
- 16.

#### 4.4 Parameters and Hyper-parameters[\[1213\]](#)

## 5 Metrics[\[1214\]](#)

| Dataset Name      | Dataset description | MS-SSIM | Perceptual loss |
|-------------------|---------------------|---------|-----------------|
| Data version zero | Training Dataset    | 0.19    | 24.34           |

## 6 Intended Use[\[1215\]](#)

|                        |                                                                                         |
|------------------------|-----------------------------------------------------------------------------------------|
| Limitations            | GPU's are required                                                                      |
| Intended Use           | It is Used to Generate Non Contrast Abdomen CT scans (No stone) for building ML models. |
| Intended User          | Radiologists                                                                            |
| Out of scope use cases | -                                                                                       |

## 7 Clinical Evaluation[\[1216\]](#)

8 Ethical considerations[\[1217\]](#)

9 Informed Consent[\[1218\]](#)

10 Legal information[\[1219\]](#)