

# FinalAssignment\_S1155135359

May 2, 2020

## 1 Keypoint Detection and Image Generation

HUANG Hejun (s1155135359)

- Presentation: 6:30pm, May 4. Each student has 2-3 minutes.

## 2 Data

```
[84]: from google.colab import drive
drive.mount('/content/gdrive/')
ROOT_FOLDER = './gdrive/My Drive/Colab Notebooks/MAEG5735-2020-Assignment4/'

import glob
print('\nContents in the data folder:')
for x in glob.glob(ROOT_FOLDER+'data/*'):
    print(x)
```

Drive already mounted at /content/gdrive/; to attempt to forcibly remount, call drive.mount("/content/gdrive/", force\_remount=True).

Contents in the data folder:

```
./gdrive/My Drive/Colab Notebooks/MAEG5735-2020-Assignment4/data/imgs1.npy
./gdrive/My Drive/Colab Notebooks/MAEG5735-2020-Assignment4/data/imgs2.npy
./gdrive/My Drive/Colab Notebooks/MAEG5735-2020-Assignment4/data/kpts3.npy
./gdrive/My Drive/Colab Notebooks/MAEG5735-2020-Assignment4/data/kpts1.npy
./gdrive/My Drive/Colab Notebooks/MAEG5735-2020-Assignment4/data/README.txt
```

There should be 4 files: \* imgs1.npy \* kpts1.npy \* imgs2.npy \* kpts3.npy

```
[85]: import matplotlib.pyplot as plt
import numpy as np

def draw_points(image, kpts):
    plt.figure()
    plt.imshow(image, cmap='gray')
    keypoints = (kpts+0.5)*IMG_SIZE
    plt.scatter(keypoints[:, 0], keypoints[:, 1], s=50, marker='.', c='r')
```

```

# load the data
IMG_SIZE = 200
IMG_TRAIN = np.load(ROOT_FOLDER+'data/imgs1.npy')
IMG_TRAIN = np.unpackbits(IMG_TRAIN).reshape((-1,IMG_SIZE,IMG_SIZE))
KPT_TRAIN = np.load(ROOT_FOLDER+'data/kpts1.npy')/IMG_SIZE - 0.5

IMG_TEST = np.load(ROOT_FOLDER+'data/imgs2.npy')
IMG_TEST = np.unpackbits(IMG_TEST).reshape((-1,IMG_SIZE,IMG_SIZE))

KPT_TEST = np.load(ROOT_FOLDER+'data/kpts3.npy')/IMG_SIZE - 0.5

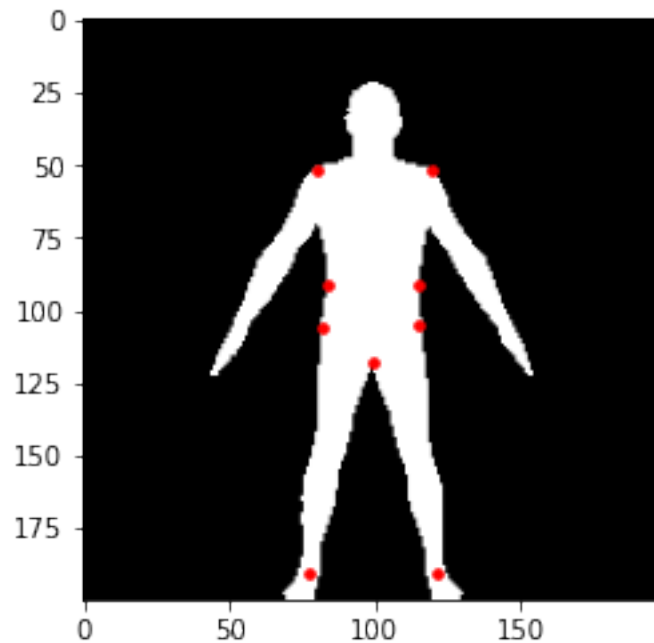
# show one
idx = 10
draw_points(IMG_TRAIN[idx,:,:], KPT_TRAIN[idx,:,:])

print(np.shape(IMG_TRAIN))
print(np.shape(KPT_TRAIN))

```

(1000, 200, 200)

(1000, 9, 2)



## 3 Task 1: Keypoint Detection

### 3.1 Keypoints Dataset loader

```
[0]: import torch
from torch.utils.data import Dataset

class KeypointsDataset(Dataset):
    '''Keypoints Dataset'''
    def __init__(self, img, kpt, train=True, transform=None):
        self.img = img
        self.kpt = kpt
        self.train = train
        self.transform = transform

    def __len__(self):
        return self.img.shape[0]

    def __getitem__(self, idx):
        image = self.img[idx,:,:].astype(np.float32)
        if self.train:
            keypoints = self.kpt[idx,:,:].ravel().astype(np.float32)
        else:
            keypoints = None
        sample = {'image': image, 'keypoints': keypoints}
        if self.transform:
            sample = self.transform(sample)
        return sample

[0]: from torch.utils.data.sampler import SubsetRandomSampler

def prepare_train_valid_loaders(trainset, valid_size=0.2,
                                batch_size=128):
    '''
    Split trainset data and prepare DataLoader for training and validation

    Args:
        trainset (Dataset): data
        valid_size (float): validation size, default=0.2
        batch_size (int) : batch size, default=128
    '''

    # obtain training indices that will be used for validation
    num_train = len(trainset)
    indices = list(range(num_train))
    np.random.shuffle(indices)
    split = int(np.floor(valid_size * num_train))
    train_idx, valid_idx = indices[split:], indices[:split]
```

```

# define samplers for obtaining training and validation batches
train_sampler = SubsetRandomSampler(train_idx)
valid_sampler = SubsetRandomSampler(valid_idx)

# prepare data loaders
train_loader = torch.utils.data.DataLoader(trainset, batch_size=batch_size,
→sampler=train_sampler)
valid_loader = torch.utils.data.DataLoader(trainset, batch_size=batch_size,
→sampler=valid_sampler)

return train_loader, valid_loader

```

```

[0]: from torchvision import transforms
import cv2

class Rescale(object):
    def __init__(self, output_size):
        assert isinstance(output_size, (int, tuple))
        self.output_size = output_size

    def __call__(self, sample):
        image, key_pts = sample['image'], sample['keypoints']
        h, w = image.shape[:2]
        new_w = np.random.randint(w, self.output_size)
        new_h = new_w
        new_h, new_w = int(new_h), int(new_w)
        img = cv2.resize(image, (new_w, new_h))
        if key_pts is not None:
            return {'image': img, 'keypoints': key_pts}
        else:
            return {'image': img}

class RandomCrop(object):
    def __init__(self, output_size):
        assert isinstance(output_size, (int, tuple))
        if isinstance(output_size, int):
            self.output_size = (output_size, output_size)
        else:
            assert len(output_size) == 2
            self.output_size = output_size

    def __call__(self, sample):
        image, key_pts = sample['image'], sample['keypoints']
        h, w = image.shape[:2]
        new_h, new_w = self.output_size
        if h == new_h:

```

```

        return sample
    top = np.random.randint(0, h - new_h)
    left = np.random.randint(0, w - new_w)
    #left = top # temp
    image = image[top: top + new_h,
                  left: left + new_w]
    if key_pts is not None:
        #key_pts = key_pts - [left/output_size, top/output_size]
        key_pts[0::2] = ((key_pts[0::2]+0.5)*w-left)/new_w-0.5
        key_pts[1::2] = ((key_pts[1::2]+0.5)*h-top)/new_h-0.5
        return {'image': image, 'keypoints': key_pts}
    else:
        return {'image': image}

class ToTensor(object):
    '''Convert ndarrays in sample to Tensors.'''
    def __call__(self, sample):
        image, keypoints = sample['image'], sample['keypoints']
        # swap color axis because
        # numpy image: H x W x C
        # torch image: C x H x W
        image = image.reshape(1, IMG_SIZE, IMG_SIZE)
        image = torch.from_numpy(image)
        if keypoints is not None:
            keypoints = torch.from_numpy(keypoints)
            return {'image': image, 'keypoints': keypoints}
        else:
            return {'image': image}

```

```

[89]: batch_size = 64
      valid_size = 0.3 # percentage of training set to use as validation

      # Define a transform to normalize the data
      tsfm_train = transforms.Compose([Rescale(205), RandomCrop(200), ToTensor()])
      tsfm_test = transforms.Compose([ToTensor()])

      print(KPT_TRAIN.shape)
      # Load the training data and test data of FRONT view
      trainset_front = KeypointsDataset(IMG_TRAIN, KPT_TRAIN, transform=tsfm_train)
      testset_front = KeypointsDataset(IMG_TEST, None, train=False,
      ↪transform=tsfm_test)

      # prepare data loaders for front view
      train_loader_front, valid_loader_front =
      ↪prepare_train_valid_loaders(trainset_front, valid_size, batch_size)
      test_loader_front = torch.utils.data.DataLoader(testset_front,
      ↪batch_size=batch_size)

```

(1000, 9, 2)

## 3.2 Prediction model setting

### 3.2.1 MLP model

```
[0]: from torch import nn, optim
import torch.nn.functional as F

class MLP(nn.Module):
    def __init__(self, input_size, output_size, hidden_layers, drop_p =0.5):
        '''
        Buid a forward network with arbitrary hidden layers.
        Arguments
        -----
        input_size (integer): size of the input layer
        output_size (integer): size of the output layer
        hidden_layers (list of integers):, the sizes of each hidden layers
        '''
        super(MLP, self).__init__()
        # hidden layers
        layer_sizes = [(input_size, hidden_layers[0])] \
            + list(zip(hidden_layers[:-1], hidden_layers[1:]))
        self.hidden_layers = nn.ModuleList([nn.Linear(h1, h2) for h1, h2 in
        ↪layer_sizes])
        self.output = nn.Linear(hidden_layers[-1], output_size)
        self.dropout = nn.Dropout(drop_p)

    def forward(self, x):
        ''' Forward pass through the network, returns the output logits '''
        # flatten inputs
        x = x.view(x.shape[0], -1)
        for layer in self.hidden_layers:
            x = F.relu(layer(x))
            x = self.dropout(x)
        x = self.output(x)
        return x
```

```
[0]: def train(train_loader, valid_loader, model, criterion, optimizer,
            n_epochs=50, saved_model='model.pt'):
    '''
    Train the model

    Args:
        train_loader (DataLoader): DataLoader for train Dataset
        valid_loader (DataLoader): DataLoader for valid Dataset
        model (nn.Module): model to be trained on
        criterion (torch.nn): loss funtion
```

```

    optimizer (torch.optim): optimization algorithms
    n_epochs (int): number of epochs to train the model
    saved_model (str): file path for saving model

Return:
    tuple of train_losses, valid_losses
'''

# initialize tracker for minimum validation loss
valid_loss_min = np.Inf # set initial "min" to infinity

train_losses = []
valid_losses = []

for epoch in range(n_epochs):
    # monitor training loss
    train_loss = 0.0
    valid_loss = 0.0

    #####
    # train the model #
    #####
    model.train() # prep model for training
    for batch in train_loader:
        # clear the gradients of all optimized variables
        optimizer.zero_grad()
        # forward pass: compute predicted outputs by passing inputs to the
→model
        output = model(batch['image'].to(device))
        # calculate the loss
        loss = criterion(output, batch['keypoints'].to(device))
        # backward pass: compute gradient of the loss with respect to model
→parameters
        loss.backward()
        # perform a single optimization step (parameter update)
        optimizer.step()
        # update running training loss
        train_loss += loss.item()*batch['image'].size(0)

    #####
    # validate the model #
    #####
    model.eval() # prep model for evaluation
    for batch in valid_loader:
        # forward pass: compute predicted outputs by passing inputs to the
→model
        output = model(batch['image'].to(device))

```

```

        # calculate the loss
        loss = criterion(output, batch['keypoints'].to(device))
        # update running validation loss
        valid_loss += loss.item()*batch['image'].size(0)

    # print training/validation statistics
    # calculate average Root Mean Square loss over an epoch
    train_loss = np.sqrt(train_loss/len(train_loader.sampler.indices))
    valid_loss = np.sqrt(valid_loss/len(valid_loader.sampler.indices))

    train_losses.append(train_loss)
    valid_losses.append(valid_loss)

    print('Epoch: {} \tTraining Loss: {:.6f} \tValidation Loss: {:.6f}'
          .format(epoch+1, train_loss, valid_loss))

    # save model if validation loss has decreased
    if valid_loss <= valid_loss_min:
        print('Validation loss decreased ({:.6f} --> {:.6f}). Saving model_
→...'
            .format(valid_loss_min, valid_loss))
        torch.save(model.state_dict(), saved_model)
        model_temp = model.state_dict()
        valid_loss_min = valid_loss
    return train_losses, valid_losses

```

```

[0]: from torch import optim
device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')

def predict(data_loader, model):
    '''
    Predict keypoints
    Args:
        data_loader (DataLoader): DataLoader for Dataset
        model (nn.Module): trained model for prediction.
    Return:
        predictions (array-like): keypoints in float (no. of images  $x_i$ 
→keypoints).
    '''

    model.eval() # prep model for evaluation

    with torch.no_grad():
        for i, batch in enumerate(data_loader):
            # forward pass: compute predicted outputs by passing inputs to the
→model
            output = model(batch['image'].to(device)).cpu().numpy()

```



```

        if i == 0:
            predictions = output
        else:
            predictions = np.vstack((predictions, output))

    return predictions

```

### 3.2.2 CNN model

```

[0]: class CNN(nn.Module):
    def __init__(self, output_size):
        super(CNN, self).__init__()
        # 200 x 200
        self.conv1 = nn.Conv2d(1, 32, 5, padding=2)
        #  $(w-f)/s+1 = 200$ 
        self.pool1 = nn.MaxPool2d(4, 4)
        # 50
        self.conv2 = nn.Conv2d(32, 64, 3, padding=2)
        #  $(54-3)/1 + 1 = 52$ 
        self.pool2 = nn.MaxPool2d(2, 2)
        # 26
        self.conv3 = nn.Conv2d(64, 128, 3)
        #  $(26-3)/1 + 1 = 24$ 
        self.pool3 = nn.MaxPool2d(2, 2)
        # 12
        self.conv4 = nn.Conv2d(128, 256, 3, stride=2)
        #  $(12-3)/2 + 1 = 5$ 
        self.conv5 = nn.Conv2d(256, 512, 1)
        #  $(5-1)/1+1 = 5$ 

        # Fully Connected Layer
        self.fc1 = nn.Linear(512*5*5, 1024)
        self.fc2 = nn.Linear(1024, output_size)
        self.drop1 = nn.Dropout(p=0.1)
        self.drop2 = nn.Dropout(p=0.25)
        self.drop3 = nn.Dropout(p=0.25)
        self.drop4 = nn.Dropout(p=0.25)
        self.drop5 = nn.Dropout(p=0.35)
        self.drop6 = nn.Dropout(p=0.4)

    def forward(self, x):
        x = self.pool1(F.relu(self.conv1(x)))
        x = self.drop1(x)
        x = self.pool2(F.relu(self.conv2(x)))
        x = self.drop2(x)
        x = self.pool3(F.relu(self.conv3(x)))
        x = self.drop3(x)

```

```

x = F.relu(self.conv4(x))
x = self.drop4(x)
x = F.relu(self.conv5(x))
x = self.drop5(x)
x = x.view(x.size(0), -1)
x = F.relu(self.fc1(x))
x = self.drop6(x)
x = self.fc2(x)
return x

```

### 3.3 Train the model using CNN

The following parameters are chosen:

- Batch\_size: 64
- Valid split: 3/7
- Learning rate: 0.0005
- Number of epochs: 220
- With 5 conv-network and 2 full connect network and the matched parameters are shown above.

```

[94]: model_front = CNN(output_size=18)
model_front = model_front.to(device)
criterion = nn.MSELoss()
optimizer = optim.Adam(model_front.parameters(), lr=0.0001)

train_losses, valid_losses = train(train_loader_front, valid_loader_front,
                                   model_front, criterion, optimizer,
                                   n_epochs=500,
                                   saved_model=ROOT_FOLDER+'model_front.pt')

```

```

Epoch: 1      Training Loss: 0.090174      Validation Loss: 0.024724
Validation loss decreased (inf --> 0.024724). Saving model ...
Epoch: 2      Training Loss: 0.041324      Validation Loss: 0.014640
Validation loss decreased (0.024724 --> 0.014640). Saving model ...
Epoch: 3      Training Loss: 0.036031      Validation Loss: 0.016622
Epoch: 4      Training Loss: 0.033536      Validation Loss: 0.013995
Validation loss decreased (0.014640 --> 0.013995). Saving model ...
Epoch: 5      Training Loss: 0.032273      Validation Loss: 0.015669
Epoch: 6      Training Loss: 0.031027      Validation Loss: 0.013902
Validation loss decreased (0.013995 --> 0.013902). Saving model ...
Epoch: 7      Training Loss: 0.030036      Validation Loss: 0.015097
Epoch: 8      Training Loss: 0.029699      Validation Loss: 0.013928
Epoch: 9      Training Loss: 0.029263      Validation Loss: 0.014426
Epoch: 10     Training Loss: 0.028350      Validation Loss: 0.013540
Validation loss decreased (0.013902 --> 0.013540). Saving model ...
Epoch: 11     Training Loss: 0.028272      Validation Loss: 0.014169

```

Epoch: 12	Training Loss: 0.027828	Validation Loss: 0.014340
Epoch: 13	Training Loss: 0.027156	Validation Loss: 0.013575
Epoch: 14	Training Loss: 0.026840	Validation Loss: 0.013410
Validation loss decreased (0.013540 --> 0.013410). Saving model ...		
Epoch: 15	Training Loss: 0.026603	Validation Loss: 0.013780
Epoch: 16	Training Loss: 0.025958	Validation Loss: 0.012912
Validation loss decreased (0.013410 --> 0.012912). Saving model ...		
Epoch: 17	Training Loss: 0.026353	Validation Loss: 0.012953
Epoch: 18	Training Loss: 0.025388	Validation Loss: 0.013403
Epoch: 19	Training Loss: 0.024885	Validation Loss: 0.012703
Validation loss decreased (0.012912 --> 0.012703). Saving model ...		
Epoch: 20	Training Loss: 0.024356	Validation Loss: 0.012414
Validation loss decreased (0.012703 --> 0.012414). Saving model ...		
Epoch: 21	Training Loss: 0.024120	Validation Loss: 0.013011
Epoch: 22	Training Loss: 0.023591	Validation Loss: 0.010592
Validation loss decreased (0.012414 --> 0.010592). Saving model ...		
Epoch: 23	Training Loss: 0.022983	Validation Loss: 0.010861
Epoch: 24	Training Loss: 0.022277	Validation Loss: 0.012299
Epoch: 25	Training Loss: 0.021718	Validation Loss: 0.008491
Validation loss decreased (0.010592 --> 0.008491). Saving model ...		
Epoch: 26	Training Loss: 0.021389	Validation Loss: 0.009709
Epoch: 27	Training Loss: 0.021387	Validation Loss: 0.010069
Epoch: 28	Training Loss: 0.020779	Validation Loss: 0.008425
Validation loss decreased (0.008491 --> 0.008425). Saving model ...		
Epoch: 29	Training Loss: 0.021055	Validation Loss: 0.011009
Epoch: 30	Training Loss: 0.020276	Validation Loss: 0.008334
Validation loss decreased (0.008425 --> 0.008334). Saving model ...		
Epoch: 31	Training Loss: 0.020951	Validation Loss: 0.008258
Validation loss decreased (0.008334 --> 0.008258). Saving model ...		
Epoch: 32	Training Loss: 0.020537	Validation Loss: 0.008800
Epoch: 33	Training Loss: 0.019983	Validation Loss: 0.007510
Validation loss decreased (0.008258 --> 0.007510). Saving model ...		
Epoch: 34	Training Loss: 0.019710	Validation Loss: 0.008639
Epoch: 35	Training Loss: 0.019834	Validation Loss: 0.007821
Epoch: 36	Training Loss: 0.019788	Validation Loss: 0.007805
Epoch: 37	Training Loss: 0.019807	Validation Loss: 0.009757
Epoch: 38	Training Loss: 0.019370	Validation Loss: 0.007758
Epoch: 39	Training Loss: 0.018948	Validation Loss: 0.007756
Epoch: 40	Training Loss: 0.019340	Validation Loss: 0.007457
Validation loss decreased (0.007510 --> 0.007457). Saving model ...		
Epoch: 41	Training Loss: 0.018778	Validation Loss: 0.007302
Validation loss decreased (0.007457 --> 0.007302). Saving model ...		
Epoch: 42	Training Loss: 0.018204	Validation Loss: 0.007117
Validation loss decreased (0.007302 --> 0.007117). Saving model ...		
Epoch: 43	Training Loss: 0.018493	Validation Loss: 0.007688
Epoch: 44	Training Loss: 0.018305	Validation Loss: 0.007261
Epoch: 45	Training Loss: 0.018224	Validation Loss: 0.007487
Epoch: 46	Training Loss: 0.018551	Validation Loss: 0.009125

Epoch: 47	Training Loss: 0.018309	Validation Loss: 0.007647
Epoch: 48	Training Loss: 0.018042	Validation Loss: 0.007923
Epoch: 49	Training Loss: 0.017956	Validation Loss: 0.007042
Validation loss decreased (0.007117 --> 0.007042). Saving model ...		
Epoch: 50	Training Loss: 0.017384	Validation Loss: 0.007827
Epoch: 51	Training Loss: 0.017352	Validation Loss: 0.007182
Epoch: 52	Training Loss: 0.017566	Validation Loss: 0.007119
Epoch: 53	Training Loss: 0.017133	Validation Loss: 0.007253
Epoch: 54	Training Loss: 0.017284	Validation Loss: 0.007198
Epoch: 55	Training Loss: 0.016785	Validation Loss: 0.006772
Validation loss decreased (0.007042 --> 0.006772). Saving model ...		
Epoch: 56	Training Loss: 0.017025	Validation Loss: 0.007095
Epoch: 57	Training Loss: 0.016674	Validation Loss: 0.007663
Epoch: 58	Training Loss: 0.016505	Validation Loss: 0.006970
Epoch: 59	Training Loss: 0.016634	Validation Loss: 0.007093
Epoch: 60	Training Loss: 0.016583	Validation Loss: 0.008366
Epoch: 61	Training Loss: 0.017272	Validation Loss: 0.007711
Epoch: 62	Training Loss: 0.016525	Validation Loss: 0.006853
Epoch: 63	Training Loss: 0.016504	Validation Loss: 0.007467
Epoch: 64	Training Loss: 0.016168	Validation Loss: 0.006736
Validation loss decreased (0.006772 --> 0.006736). Saving model ...		
Epoch: 65	Training Loss: 0.016652	Validation Loss: 0.006978
Epoch: 66	Training Loss: 0.016557	Validation Loss: 0.006973
Epoch: 67	Training Loss: 0.016826	Validation Loss: 0.008958
Epoch: 68	Training Loss: 0.016022	Validation Loss: 0.007444
Epoch: 69	Training Loss: 0.015627	Validation Loss: 0.006779
Epoch: 70	Training Loss: 0.016162	Validation Loss: 0.006627
Validation loss decreased (0.006736 --> 0.006627). Saving model ...		
Epoch: 71	Training Loss: 0.015883	Validation Loss: 0.006797
Epoch: 72	Training Loss: 0.015695	Validation Loss: 0.006746
Epoch: 73	Training Loss: 0.015656	Validation Loss: 0.007973
Epoch: 74	Training Loss: 0.015454	Validation Loss: 0.006728
Epoch: 75	Training Loss: 0.015199	Validation Loss: 0.006793
Epoch: 76	Training Loss: 0.015621	Validation Loss: 0.006481
Validation loss decreased (0.006627 --> 0.006481). Saving model ...		
Epoch: 77	Training Loss: 0.015044	Validation Loss: 0.007498
Epoch: 78	Training Loss: 0.014969	Validation Loss: 0.008367
Epoch: 79	Training Loss: 0.015354	Validation Loss: 0.006917
Epoch: 80	Training Loss: 0.015635	Validation Loss: 0.007364
Epoch: 81	Training Loss: 0.015357	Validation Loss: 0.006573
Epoch: 82	Training Loss: 0.015190	Validation Loss: 0.006295
Validation loss decreased (0.006481 --> 0.006295). Saving model ...		
Epoch: 83	Training Loss: 0.014701	Validation Loss: 0.006500
Epoch: 84	Training Loss: 0.015043	Validation Loss: 0.006536
Epoch: 85	Training Loss: 0.014860	Validation Loss: 0.006187
Validation loss decreased (0.006295 --> 0.006187). Saving model ...		
Epoch: 86	Training Loss: 0.014540	Validation Loss: 0.006923
Epoch: 87	Training Loss: 0.014908	Validation Loss: 0.006545

Epoch: 88	Training Loss: 0.014466	Validation Loss: 0.006841
Epoch: 89	Training Loss: 0.014617	Validation Loss: 0.007949
Epoch: 90	Training Loss: 0.014834	Validation Loss: 0.006223
Epoch: 91	Training Loss: 0.014633	Validation Loss: 0.007552
Epoch: 92	Training Loss: 0.015067	Validation Loss: 0.006965
Epoch: 93	Training Loss: 0.014163	Validation Loss: 0.007690
Epoch: 94	Training Loss: 0.014347	Validation Loss: 0.006012
Validation loss decreased (0.006187 --> 0.006012). Saving model ...		
Epoch: 95	Training Loss: 0.014473	Validation Loss: 0.006184
Epoch: 96	Training Loss: 0.013646	Validation Loss: 0.006463
Epoch: 97	Training Loss: 0.014236	Validation Loss: 0.005972
Validation loss decreased (0.006012 --> 0.005972). Saving model ...		
Epoch: 98	Training Loss: 0.013677	Validation Loss: 0.007576
Epoch: 99	Training Loss: 0.014286	Validation Loss: 0.005847
Validation loss decreased (0.005972 --> 0.005847). Saving model ...		
Epoch: 100	Training Loss: 0.014247	Validation Loss: 0.006224
Epoch: 101	Training Loss: 0.013756	Validation Loss: 0.005935
Epoch: 102	Training Loss: 0.013898	Validation Loss: 0.007692
Epoch: 103	Training Loss: 0.013983	Validation Loss: 0.006247
Epoch: 104	Training Loss: 0.013915	Validation Loss: 0.006381
Epoch: 105	Training Loss: 0.013808	Validation Loss: 0.006087
Epoch: 106	Training Loss: 0.013574	Validation Loss: 0.007065
Epoch: 107	Training Loss: 0.014452	Validation Loss: 0.007322
Epoch: 108	Training Loss: 0.013241	Validation Loss: 0.005918
Epoch: 109	Training Loss: 0.013938	Validation Loss: 0.005767
Validation loss decreased (0.005847 --> 0.005767). Saving model ...		
Epoch: 110	Training Loss: 0.013490	Validation Loss: 0.005709
Validation loss decreased (0.005767 --> 0.005709). Saving model ...		
Epoch: 111	Training Loss: 0.013961	Validation Loss: 0.005880
Epoch: 112	Training Loss: 0.013469	Validation Loss: 0.005644
Validation loss decreased (0.005709 --> 0.005644). Saving model ...		
Epoch: 113	Training Loss: 0.013574	Validation Loss: 0.006866
Epoch: 114	Training Loss: 0.013797	Validation Loss: 0.005664
Epoch: 115	Training Loss: 0.013583	Validation Loss: 0.007513
Epoch: 116	Training Loss: 0.013691	Validation Loss: 0.006337
Epoch: 117	Training Loss: 0.013136	Validation Loss: 0.005665
Epoch: 118	Training Loss: 0.013377	Validation Loss: 0.006067
Epoch: 119	Training Loss: 0.012863	Validation Loss: 0.006009
Epoch: 120	Training Loss: 0.013676	Validation Loss: 0.006095
Epoch: 121	Training Loss: 0.013404	Validation Loss: 0.006086
Epoch: 122	Training Loss: 0.013452	Validation Loss: 0.005819
Epoch: 123	Training Loss: 0.012807	Validation Loss: 0.009341
Epoch: 124	Training Loss: 0.013660	Validation Loss: 0.005525
Validation loss decreased (0.005644 --> 0.005525). Saving model ...		
Epoch: 125	Training Loss: 0.012806	Validation Loss: 0.005940
Epoch: 126	Training Loss: 0.013352	Validation Loss: 0.006491
Epoch: 127	Training Loss: 0.012644	Validation Loss: 0.005567
Epoch: 128	Training Loss: 0.012584	Validation Loss: 0.005334

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Validation loss decreased (0.005525 --> 0.005334). Saving model ...
Epoch: 129      Training Loss: 0.012853      Validation Loss: 0.005432
Epoch: 130      Training Loss: 0.013288      Validation Loss: 0.005301
Validation loss decreased (0.005334 --> 0.005301). Saving model ...
Epoch: 131      Training Loss: 0.012833      Validation Loss: 0.005399
Epoch: 132      Training Loss: 0.012849      Validation Loss: 0.005460
Epoch: 133      Training Loss: 0.012917      Validation Loss: 0.005996
Epoch: 134      Training Loss: 0.013406      Validation Loss: 0.006640
Epoch: 135      Training Loss: 0.012796      Validation Loss: 0.005121
Validation loss decreased (0.005301 --> 0.005121). Saving model ...
Epoch: 136      Training Loss: 0.012634      Validation Loss: 0.007272
Epoch: 137      Training Loss: 0.012909      Validation Loss: 0.005734
Epoch: 138      Training Loss: 0.012783      Validation Loss: 0.005223
Epoch: 139      Training Loss: 0.013103      Validation Loss: 0.006479
Epoch: 140      Training Loss: 0.012777      Validation Loss: 0.005235
Epoch: 141      Training Loss: 0.012592      Validation Loss: 0.006889
Epoch: 142      Training Loss: 0.012651      Validation Loss: 0.004776
Validation loss decreased (0.005121 --> 0.004776). Saving model ...
Epoch: 143      Training Loss: 0.012449      Validation Loss: 0.005242
Epoch: 144      Training Loss: 0.012125      Validation Loss: 0.006304
Epoch: 145      Training Loss: 0.011872      Validation Loss: 0.004916
Epoch: 146      Training Loss: 0.012366      Validation Loss: 0.004962
Epoch: 147      Training Loss: 0.012140      Validation Loss: 0.005127
Epoch: 148      Training Loss: 0.012338      Validation Loss: 0.005325
Epoch: 149      Training Loss: 0.012404      Validation Loss: 0.004673
Validation loss decreased (0.004776 --> 0.004673). Saving model ...
Epoch: 150      Training Loss: 0.012923      Validation Loss: 0.006817
Epoch: 151      Training Loss: 0.012327      Validation Loss: 0.005090
Epoch: 152      Training Loss: 0.011961      Validation Loss: 0.005793
Epoch: 153      Training Loss: 0.012217      Validation Loss: 0.005190
Epoch: 154      Training Loss: 0.012459      Validation Loss: 0.004946
Epoch: 155      Training Loss: 0.012376      Validation Loss: 0.004761
Epoch: 156      Training Loss: 0.012261      Validation Loss: 0.005728
Epoch: 157      Training Loss: 0.012535      Validation Loss: 0.007615
Epoch: 158      Training Loss: 0.012407      Validation Loss: 0.005525
Epoch: 159      Training Loss: 0.012602      Validation Loss: 0.008071
Epoch: 160      Training Loss: 0.012510      Validation Loss: 0.004853
Epoch: 161      Training Loss: 0.011931      Validation Loss: 0.004837
Epoch: 162      Training Loss: 0.011866      Validation Loss: 0.005210
Epoch: 163      Training Loss: 0.012271      Validation Loss: 0.005692
Epoch: 164      Training Loss: 0.011450      Validation Loss: 0.005542
Epoch: 165      Training Loss: 0.011873      Validation Loss: 0.005047
Epoch: 166      Training Loss: 0.012351      Validation Loss: 0.004727
Epoch: 167      Training Loss: 0.011401      Validation Loss: 0.004851
Epoch: 168      Training Loss: 0.012318      Validation Loss: 0.007487
Epoch: 169      Training Loss: 0.011749      Validation Loss: 0.004705
Epoch: 170      Training Loss: 0.011877      Validation Loss: 0.004904
Epoch: 171      Training Loss: 0.011622      Validation Loss: 0.005338

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Epoch: 172	Training Loss: 0.011835	Validation Loss: 0.004643
Validation loss decreased (0.004673 --> 0.004643). Saving model ...		
Epoch: 173	Training Loss: 0.012044	Validation Loss: 0.005594
Epoch: 174	Training Loss: 0.012463	Validation Loss: 0.005663
Epoch: 175	Training Loss: 0.011348	Validation Loss: 0.005718
Epoch: 176	Training Loss: 0.011533	Validation Loss: 0.004886
Epoch: 177	Training Loss: 0.012089	Validation Loss: 0.005196
Epoch: 178	Training Loss: 0.012154	Validation Loss: 0.004798
Epoch: 179	Training Loss: 0.011965	Validation Loss: 0.007128
Epoch: 180	Training Loss: 0.010996	Validation Loss: 0.004694
Epoch: 181	Training Loss: 0.012448	Validation Loss: 0.006228
Epoch: 182	Training Loss: 0.012386	Validation Loss: 0.007564
Epoch: 183	Training Loss: 0.011686	Validation Loss: 0.005121
Epoch: 184	Training Loss: 0.011206	Validation Loss: 0.004639
Validation loss decreased (0.004643 --> 0.004639). Saving model ...		
Epoch: 185	Training Loss: 0.011232	Validation Loss: 0.004726
Epoch: 186	Training Loss: 0.011536	Validation Loss: 0.005534
Epoch: 187	Training Loss: 0.011282	Validation Loss: 0.006088
Epoch: 188	Training Loss: 0.011690	Validation Loss: 0.006342
Epoch: 189	Training Loss: 0.011357	Validation Loss: 0.004693
Epoch: 190	Training Loss: 0.011363	Validation Loss: 0.005146
Epoch: 191	Training Loss: 0.010912	Validation Loss: 0.004416
Validation loss decreased (0.004639 --> 0.004416). Saving model ...		
Epoch: 192	Training Loss: 0.011341	Validation Loss: 0.005211
Epoch: 193	Training Loss: 0.011183	Validation Loss: 0.005418
Epoch: 194	Training Loss: 0.011540	Validation Loss: 0.004732
Epoch: 195	Training Loss: 0.011186	Validation Loss: 0.004660
Epoch: 196	Training Loss: 0.011310	Validation Loss: 0.004407
Validation loss decreased (0.004416 --> 0.004407). Saving model ...		
Epoch: 197	Training Loss: 0.011149	Validation Loss: 0.005201
Epoch: 198	Training Loss: 0.011221	Validation Loss: 0.004078
Validation loss decreased (0.004407 --> 0.004078). Saving model ...		
Epoch: 199	Training Loss: 0.011132	Validation Loss: 0.004973
Epoch: 200	Training Loss: 0.012148	Validation Loss: 0.007206
Epoch: 201	Training Loss: 0.011640	Validation Loss: 0.006658
Epoch: 202	Training Loss: 0.011103	Validation Loss: 0.005051
Epoch: 203	Training Loss: 0.011498	Validation Loss: 0.004801
Epoch: 204	Training Loss: 0.011105	Validation Loss: 0.005983
Epoch: 205	Training Loss: 0.011484	Validation Loss: 0.005264
Epoch: 206	Training Loss: 0.011347	Validation Loss: 0.004462
Epoch: 207	Training Loss: 0.011391	Validation Loss: 0.004816
Epoch: 208	Training Loss: 0.010959	Validation Loss: 0.004412
Epoch: 209	Training Loss: 0.010908	Validation Loss: 0.004342
Epoch: 210	Training Loss: 0.010748	Validation Loss: 0.004335
Epoch: 211	Training Loss: 0.011365	Validation Loss: 0.007130
Epoch: 212	Training Loss: 0.011237	Validation Loss: 0.007530
Epoch: 213	Training Loss: 0.010859	Validation Loss: 0.004933
Epoch: 214	Training Loss: 0.010533	Validation Loss: 0.006997

Epoch: 215	Training Loss: 0.010971	Validation Loss: 0.004262
Epoch: 216	Training Loss: 0.011521	Validation Loss: 0.007979
Epoch: 217	Training Loss: 0.011159	Validation Loss: 0.007150
Epoch: 218	Training Loss: 0.011538	Validation Loss: 0.003885
Validation loss decreased (0.004078 --> 0.003885). Saving model ...		
Epoch: 219	Training Loss: 0.011203	Validation Loss: 0.004142
Epoch: 220	Training Loss: 0.010615	Validation Loss: 0.004941
Epoch: 221	Training Loss: 0.010799	Validation Loss: 0.005884
Epoch: 222	Training Loss: 0.010639	Validation Loss: 0.004162
Epoch: 223	Training Loss: 0.010630	Validation Loss: 0.004626
Epoch: 224	Training Loss: 0.010486	Validation Loss: 0.007393
Epoch: 225	Training Loss: 0.010964	Validation Loss: 0.004949
Epoch: 226	Training Loss: 0.010614	Validation Loss: 0.005383
Epoch: 227	Training Loss: 0.010786	Validation Loss: 0.005796
Epoch: 228	Training Loss: 0.010845	Validation Loss: 0.005446
Epoch: 229	Training Loss: 0.011058	Validation Loss: 0.004430
Epoch: 230	Training Loss: 0.011012	Validation Loss: 0.004630
Epoch: 231	Training Loss: 0.011095	Validation Loss: 0.006834
Epoch: 232	Training Loss: 0.010877	Validation Loss: 0.007265
Epoch: 233	Training Loss: 0.011292	Validation Loss: 0.004856
Epoch: 234	Training Loss: 0.012161	Validation Loss: 0.009983
Epoch: 235	Training Loss: 0.011498	Validation Loss: 0.004090
Epoch: 236	Training Loss: 0.010992	Validation Loss: 0.007590
Epoch: 237	Training Loss: 0.010499	Validation Loss: 0.005826
Epoch: 238	Training Loss: 0.010155	Validation Loss: 0.005156
Epoch: 239	Training Loss: 0.010726	Validation Loss: 0.005320
Epoch: 240	Training Loss: 0.010714	Validation Loss: 0.004335
Epoch: 241	Training Loss: 0.011117	Validation Loss: 0.009608
Epoch: 242	Training Loss: 0.011423	Validation Loss: 0.004174
Epoch: 243	Training Loss: 0.010690	Validation Loss: 0.006830
Epoch: 244	Training Loss: 0.010555	Validation Loss: 0.004530
Epoch: 245	Training Loss: 0.010909	Validation Loss: 0.005726
Epoch: 246	Training Loss: 0.010534	Validation Loss: 0.003712
Validation loss decreased (0.003885 --> 0.003712). Saving model ...		
Epoch: 247	Training Loss: 0.010724	Validation Loss: 0.009592
Epoch: 248	Training Loss: 0.010473	Validation Loss: 0.004515
Epoch: 249	Training Loss: 0.011602	Validation Loss: 0.004433
Epoch: 250	Training Loss: 0.010739	Validation Loss: 0.005368
Epoch: 251	Training Loss: 0.010218	Validation Loss: 0.004833
Epoch: 252	Training Loss: 0.010141	Validation Loss: 0.004915
Epoch: 253	Training Loss: 0.010311	Validation Loss: 0.007475
Epoch: 254	Training Loss: 0.010222	Validation Loss: 0.005745
Epoch: 255	Training Loss: 0.010516	Validation Loss: 0.005640
Epoch: 256	Training Loss: 0.010313	Validation Loss: 0.004154
Epoch: 257	Training Loss: 0.010042	Validation Loss: 0.007480
Epoch: 258	Training Loss: 0.010740	Validation Loss: 0.004375
Epoch: 259	Training Loss: 0.010617	Validation Loss: 0.006842
Epoch: 260	Training Loss: 0.010218	Validation Loss: 0.004227



Epoch: 261	Training Loss: 0.010180	Validation Loss: 0.005058
Epoch: 262	Training Loss: 0.010185	Validation Loss: 0.003923
Epoch: 263	Training Loss: 0.009803	Validation Loss: 0.006099
Epoch: 264	Training Loss: 0.010242	Validation Loss: 0.005686
Epoch: 265	Training Loss: 0.010136	Validation Loss: 0.004096
Epoch: 266	Training Loss: 0.009961	Validation Loss: 0.004604
Epoch: 267	Training Loss: 0.010235	Validation Loss: 0.005083
Epoch: 268	Training Loss: 0.010195	Validation Loss: 0.006619
Epoch: 269	Training Loss: 0.009806	Validation Loss: 0.004527
Epoch: 270	Training Loss: 0.010171	Validation Loss: 0.006489
Epoch: 271	Training Loss: 0.010424	Validation Loss: 0.005547
Epoch: 272	Training Loss: 0.009936	Validation Loss: 0.005215
Epoch: 273	Training Loss: 0.009975	Validation Loss: 0.004148
Epoch: 274	Training Loss: 0.010166	Validation Loss: 0.006285
Epoch: 275	Training Loss: 0.010463	Validation Loss: 0.004422
Epoch: 276	Training Loss: 0.010645	Validation Loss: 0.006096
Epoch: 277	Training Loss: 0.010344	Validation Loss: 0.007042
Epoch: 278	Training Loss: 0.010309	Validation Loss: 0.003828
Epoch: 279	Training Loss: 0.010151	Validation Loss: 0.004518
Epoch: 280	Training Loss: 0.010226	Validation Loss: 0.007048
Epoch: 281	Training Loss: 0.010841	Validation Loss: 0.004254
Epoch: 282	Training Loss: 0.010057	Validation Loss: 0.003773
Epoch: 283	Training Loss: 0.009887	Validation Loss: 0.006066
Epoch: 284	Training Loss: 0.010330	Validation Loss: 0.004194
Epoch: 285	Training Loss: 0.009846	Validation Loss: 0.006193
Epoch: 286	Training Loss: 0.010105	Validation Loss: 0.003612
Validation loss decreased (0.003712 --> 0.003612). Saving model ...		
Epoch: 287	Training Loss: 0.009944	Validation Loss: 0.003953
Epoch: 288	Training Loss: 0.009852	Validation Loss: 0.005473
Epoch: 289	Training Loss: 0.010121	Validation Loss: 0.006094
Epoch: 290	Training Loss: 0.010480	Validation Loss: 0.005064
Epoch: 291	Training Loss: 0.010115	Validation Loss: 0.006049
Epoch: 292	Training Loss: 0.010510	Validation Loss: 0.004988
Epoch: 293	Training Loss: 0.010192	Validation Loss: 0.003962
Epoch: 294	Training Loss: 0.009525	Validation Loss: 0.004819
Epoch: 295	Training Loss: 0.009721	Validation Loss: 0.003797
Epoch: 296	Training Loss: 0.009460	Validation Loss: 0.004400
Epoch: 297	Training Loss: 0.010071	Validation Loss: 0.006559
Epoch: 298	Training Loss: 0.010050	Validation Loss: 0.004062
Epoch: 299	Training Loss: 0.009941	Validation Loss: 0.006218
Epoch: 300	Training Loss: 0.010170	Validation Loss: 0.004086
Epoch: 301	Training Loss: 0.009817	Validation Loss: 0.006070
Epoch: 302	Training Loss: 0.009705	Validation Loss: 0.005801
Epoch: 303	Training Loss: 0.009436	Validation Loss: 0.004760
Epoch: 304	Training Loss: 0.009870	Validation Loss: 0.004258
Epoch: 305	Training Loss: 0.009857	Validation Loss: 0.006072
Epoch: 306	Training Loss: 0.010122	Validation Loss: 0.004927
Epoch: 307	Training Loss: 0.010135	Validation Loss: 0.003436

Validation loss decreased (0.003612 --> 0.003436). Saving model ...

Epoch: 308	Training Loss: 0.009995	Validation Loss: 0.006233
Epoch: 309	Training Loss: 0.009902	Validation Loss: 0.006882
Epoch: 310	Training Loss: 0.009743	Validation Loss: 0.004729
Epoch: 311	Training Loss: 0.009896	Validation Loss: 0.008122
Epoch: 312	Training Loss: 0.010282	Validation Loss: 0.003902
Epoch: 313	Training Loss: 0.009926	Validation Loss: 0.007397
Epoch: 314	Training Loss: 0.009606	Validation Loss: 0.003604
Epoch: 315	Training Loss: 0.009541	Validation Loss: 0.007494
Epoch: 316	Training Loss: 0.009580	Validation Loss: 0.005361
Epoch: 317	Training Loss: 0.009602	Validation Loss: 0.004624
Epoch: 318	Training Loss: 0.010287	Validation Loss: 0.009856
Epoch: 319	Training Loss: 0.009863	Validation Loss: 0.004906
Epoch: 320	Training Loss: 0.009228	Validation Loss: 0.005385
Epoch: 321	Training Loss: 0.009808	Validation Loss: 0.007297
Epoch: 322	Training Loss: 0.009878	Validation Loss: 0.004398
Epoch: 323	Training Loss: 0.009853	Validation Loss: 0.004129
Epoch: 324	Training Loss: 0.009428	Validation Loss: 0.003685
Epoch: 325	Training Loss: 0.009444	Validation Loss: 0.003740
Epoch: 326	Training Loss: 0.009870	Validation Loss: 0.008396
Epoch: 327	Training Loss: 0.009916	Validation Loss: 0.005017
Epoch: 328	Training Loss: 0.009245	Validation Loss: 0.007278
Epoch: 329	Training Loss: 0.009556	Validation Loss: 0.004679
Epoch: 330	Training Loss: 0.009717	Validation Loss: 0.004890
Epoch: 331	Training Loss: 0.009794	Validation Loss: 0.006080
Epoch: 332	Training Loss: 0.009488	Validation Loss: 0.006454
Epoch: 333	Training Loss: 0.009616	Validation Loss: 0.006243
Epoch: 334	Training Loss: 0.009537	Validation Loss: 0.004327
Epoch: 335	Training Loss: 0.009368	Validation Loss: 0.005644
Epoch: 336	Training Loss: 0.009267	Validation Loss: 0.006527
Epoch: 337	Training Loss: 0.009786	Validation Loss: 0.003556
Epoch: 338	Training Loss: 0.009951	Validation Loss: 0.003682
Epoch: 339	Training Loss: 0.009381	Validation Loss: 0.006022
Epoch: 340	Training Loss: 0.009676	Validation Loss: 0.004487
Epoch: 341	Training Loss: 0.009758	Validation Loss: 0.003575
Epoch: 342	Training Loss: 0.010685	Validation Loss: 0.009163
Epoch: 343	Training Loss: 0.010113	Validation Loss: 0.003742
Epoch: 344	Training Loss: 0.009685	Validation Loss: 0.004670
Epoch: 345	Training Loss: 0.009707	Validation Loss: 0.005450
Epoch: 346	Training Loss: 0.009662	Validation Loss: 0.004033
Epoch: 347	Training Loss: 0.009598	Validation Loss: 0.006523
Epoch: 348	Training Loss: 0.008929	Validation Loss: 0.004687
Epoch: 349	Training Loss: 0.009131	Validation Loss: 0.005113
Epoch: 350	Training Loss: 0.009249	Validation Loss: 0.003805
Epoch: 351	Training Loss: 0.009078	Validation Loss: 0.003828
Epoch: 352	Training Loss: 0.008979	Validation Loss: 0.004827
Epoch: 353	Training Loss: 0.009380	Validation Loss: 0.005859
Epoch: 354	Training Loss: 0.009415	Validation Loss: 0.004830

Epoch: 355	Training Loss: 0.009642	Validation Loss: 0.004005
Epoch: 356	Training Loss: 0.009613	Validation Loss: 0.006443
Epoch: 357	Training Loss: 0.009394	Validation Loss: 0.004462
Epoch: 358	Training Loss: 0.009264	Validation Loss: 0.004705
Epoch: 359	Training Loss: 0.009556	Validation Loss: 0.006988
Epoch: 360	Training Loss: 0.009448	Validation Loss: 0.004911
Epoch: 361	Training Loss: 0.009470	Validation Loss: 0.005612
Epoch: 362	Training Loss: 0.009265	Validation Loss: 0.003406
Validation loss decreased (0.003436 --> 0.003406). Saving model ...		
Epoch: 363	Training Loss: 0.009419	Validation Loss: 0.006743
Epoch: 364	Training Loss: 0.009007	Validation Loss: 0.004938
Epoch: 365	Training Loss: 0.009174	Validation Loss: 0.006790
Epoch: 366	Training Loss: 0.009107	Validation Loss: 0.004735
Epoch: 367	Training Loss: 0.009472	Validation Loss: 0.004173
Epoch: 368	Training Loss: 0.009251	Validation Loss: 0.005986
Epoch: 369	Training Loss: 0.009043	Validation Loss: 0.004953
Epoch: 370	Training Loss: 0.009113	Validation Loss: 0.005088
Epoch: 371	Training Loss: 0.009010	Validation Loss: 0.005626
Epoch: 372	Training Loss: 0.009118	Validation Loss: 0.003384
Validation loss decreased (0.003406 --> 0.003384). Saving model ...		
Epoch: 373	Training Loss: 0.009297	Validation Loss: 0.005182
Epoch: 374	Training Loss: 0.009222	Validation Loss: 0.005053
Epoch: 375	Training Loss: 0.009381	Validation Loss: 0.005956
Epoch: 376	Training Loss: 0.008977	Validation Loss: 0.006828
Epoch: 377	Training Loss: 0.009038	Validation Loss: 0.004731
Epoch: 378	Training Loss: 0.009313	Validation Loss: 0.005340
Epoch: 379	Training Loss: 0.009004	Validation Loss: 0.006963
Epoch: 380	Training Loss: 0.010293	Validation Loss: 0.003788
Epoch: 381	Training Loss: 0.009406	Validation Loss: 0.007204
Epoch: 382	Training Loss: 0.009222	Validation Loss: 0.004090
Epoch: 383	Training Loss: 0.009044	Validation Loss: 0.007854
Epoch: 384	Training Loss: 0.009520	Validation Loss: 0.004386
Epoch: 385	Training Loss: 0.009124	Validation Loss: 0.005869
Epoch: 386	Training Loss: 0.008955	Validation Loss: 0.004407
Epoch: 387	Training Loss: 0.009164	Validation Loss: 0.004975
Epoch: 388	Training Loss: 0.009127	Validation Loss: 0.004293
Epoch: 389	Training Loss: 0.008919	Validation Loss: 0.005385
Epoch: 390	Training Loss: 0.008931	Validation Loss: 0.004568
Epoch: 391	Training Loss: 0.009145	Validation Loss: 0.008395
Epoch: 392	Training Loss: 0.009507	Validation Loss: 0.003372
Validation loss decreased (0.003384 --> 0.003372). Saving model ...		
Epoch: 393	Training Loss: 0.009060	Validation Loss: 0.008246
Epoch: 394	Training Loss: 0.009083	Validation Loss: 0.004593
Epoch: 395	Training Loss: 0.009076	Validation Loss: 0.005719
Epoch: 396	Training Loss: 0.008664	Validation Loss: 0.006119
Epoch: 397	Training Loss: 0.008872	Validation Loss: 0.004048
Epoch: 398	Training Loss: 0.008991	Validation Loss: 0.005367
Epoch: 399	Training Loss: 0.008534	Validation Loss: 0.004016

Epoch: 400	Training Loss: 0.008719	Validation Loss: 0.004427
Epoch: 401	Training Loss: 0.009233	Validation Loss: 0.003559
Epoch: 402	Training Loss: 0.008988	Validation Loss: 0.005059
Epoch: 403	Training Loss: 0.009036	Validation Loss: 0.007836
Epoch: 404	Training Loss: 0.009564	Validation Loss: 0.005335
Epoch: 405	Training Loss: 0.008670	Validation Loss: 0.003953
Epoch: 406	Training Loss: 0.008685	Validation Loss: 0.008051
Epoch: 407	Training Loss: 0.008792	Validation Loss: 0.003641
Epoch: 408	Training Loss: 0.009285	Validation Loss: 0.005144
Epoch: 409	Training Loss: 0.008806	Validation Loss: 0.005120
Epoch: 410	Training Loss: 0.008783	Validation Loss: 0.004590
Epoch: 411	Training Loss: 0.009250	Validation Loss: 0.006735
Epoch: 412	Training Loss: 0.009107	Validation Loss: 0.004467
Epoch: 413	Training Loss: 0.008992	Validation Loss: 0.003733
Epoch: 414	Training Loss: 0.008910	Validation Loss: 0.003963
Epoch: 415	Training Loss: 0.009130	Validation Loss: 0.008706
Epoch: 416	Training Loss: 0.008732	Validation Loss: 0.006350
Epoch: 417	Training Loss: 0.009055	Validation Loss: 0.004497
Epoch: 418	Training Loss: 0.008636	Validation Loss: 0.006141
Epoch: 419	Training Loss: 0.009013	Validation Loss: 0.005323
Epoch: 420	Training Loss: 0.008382	Validation Loss: 0.004411
Epoch: 421	Training Loss: 0.008747	Validation Loss: 0.005595
Epoch: 422	Training Loss: 0.008781	Validation Loss: 0.003723
Epoch: 423	Training Loss: 0.008988	Validation Loss: 0.004199
Epoch: 424	Training Loss: 0.008564	Validation Loss: 0.007167
Epoch: 425	Training Loss: 0.008676	Validation Loss: 0.006697
Epoch: 426	Training Loss: 0.008733	Validation Loss: 0.006208
Epoch: 427	Training Loss: 0.008377	Validation Loss: 0.005975
Epoch: 428	Training Loss: 0.008717	Validation Loss: 0.005519
Epoch: 429	Training Loss: 0.009004	Validation Loss: 0.003588
Epoch: 430	Training Loss: 0.009132	Validation Loss: 0.008587
Epoch: 431	Training Loss: 0.009103	Validation Loss: 0.004324
Epoch: 432	Training Loss: 0.008635	Validation Loss: 0.007321
Epoch: 433	Training Loss: 0.009008	Validation Loss: 0.004315
Epoch: 434	Training Loss: 0.008575	Validation Loss: 0.004161
Epoch: 435	Training Loss: 0.008905	Validation Loss: 0.004485
Epoch: 436	Training Loss: 0.008820	Validation Loss: 0.007955
Epoch: 437	Training Loss: 0.008925	Validation Loss: 0.006659
Epoch: 438	Training Loss: 0.008657	Validation Loss: 0.004735
Epoch: 439	Training Loss: 0.008911	Validation Loss: 0.006945
Epoch: 440	Training Loss: 0.009082	Validation Loss: 0.006913
Epoch: 441	Training Loss: 0.008927	Validation Loss: 0.003268
Validation loss decreased (0.003372 --> 0.003268). Saving model ...		
Epoch: 442	Training Loss: 0.008747	Validation Loss: 0.004709
Epoch: 443	Training Loss: 0.008813	Validation Loss: 0.005770
Epoch: 444	Training Loss: 0.008617	Validation Loss: 0.003785
Epoch: 445	Training Loss: 0.008868	Validation Loss: 0.007179
Epoch: 446	Training Loss: 0.008368	Validation Loss: 0.004114

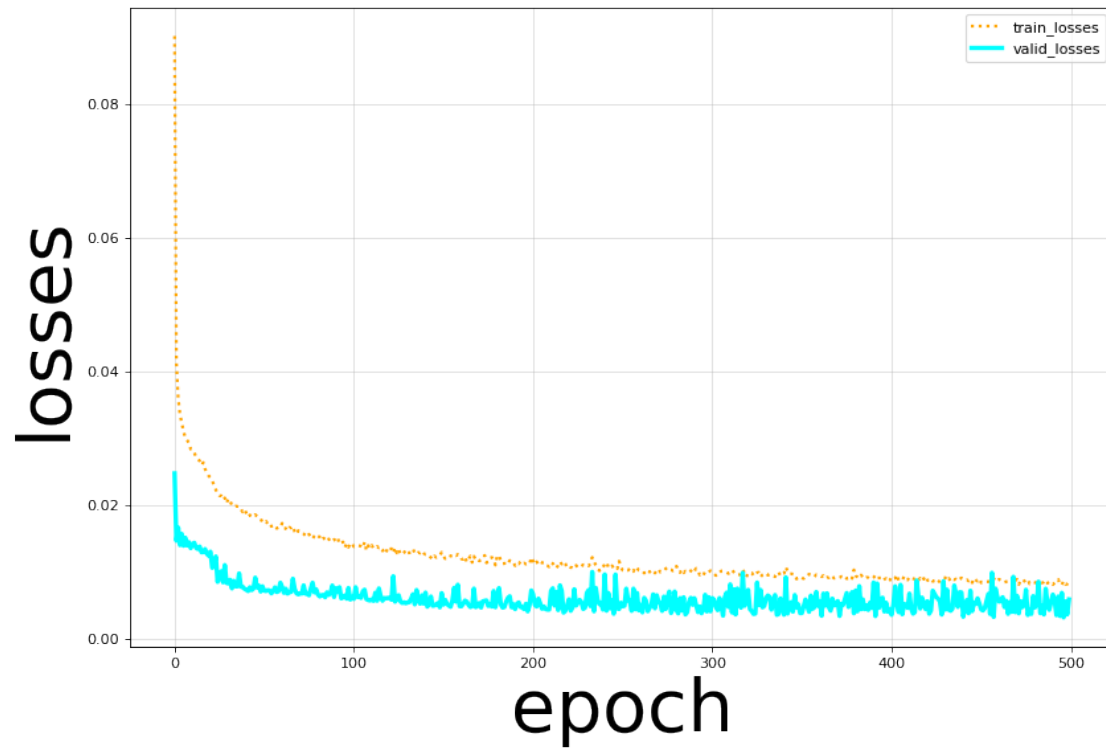
Epoch: 447	Training Loss: 0.009047	Validation Loss: 0.005411
Epoch: 448	Training Loss: 0.008784	Validation Loss: 0.005185
Epoch: 449	Training Loss: 0.008592	Validation Loss: 0.006026
Epoch: 450	Training Loss: 0.008934	Validation Loss: 0.005252
Epoch: 451	Training Loss: 0.008361	Validation Loss: 0.004367
Epoch: 452	Training Loss: 0.008424	Validation Loss: 0.004202
Epoch: 453	Training Loss: 0.008503	Validation Loss: 0.004560
Epoch: 454	Training Loss: 0.008525	Validation Loss: 0.005656
Epoch: 455	Training Loss: 0.008519	Validation Loss: 0.006674
Epoch: 456	Training Loss: 0.008608	Validation Loss: 0.003695
Epoch: 457	Training Loss: 0.009016	Validation Loss: 0.009855
Epoch: 458	Training Loss: 0.008880	Validation Loss: 0.003246
Validation loss decreased (0.003268 --> 0.003246). Saving model ...		
Epoch: 459	Training Loss: 0.009229	Validation Loss: 0.007493
Epoch: 460	Training Loss: 0.008814	Validation Loss: 0.006278
Epoch: 461	Training Loss: 0.008883	Validation Loss: 0.005536
Epoch: 462	Training Loss: 0.008701	Validation Loss: 0.005603
Epoch: 463	Training Loss: 0.008656	Validation Loss: 0.006202
Epoch: 464	Training Loss: 0.008419	Validation Loss: 0.004720
Epoch: 465	Training Loss: 0.008530	Validation Loss: 0.005669
Epoch: 466	Training Loss: 0.008611	Validation Loss: 0.003700
Epoch: 467	Training Loss: 0.008670	Validation Loss: 0.005445
Epoch: 468	Training Loss: 0.008891	Validation Loss: 0.003987
Epoch: 469	Training Loss: 0.009331	Validation Loss: 0.009247
Epoch: 470	Training Loss: 0.008692	Validation Loss: 0.003951
Epoch: 471	Training Loss: 0.008877	Validation Loss: 0.006406
Epoch: 472	Training Loss: 0.008105	Validation Loss: 0.003774
Epoch: 473	Training Loss: 0.008687	Validation Loss: 0.005624
Epoch: 474	Training Loss: 0.008428	Validation Loss: 0.004717
Epoch: 475	Training Loss: 0.008609	Validation Loss: 0.005044
Epoch: 476	Training Loss: 0.008749	Validation Loss: 0.007316
Epoch: 477	Training Loss: 0.008932	Validation Loss: 0.003534
Epoch: 478	Training Loss: 0.008211	Validation Loss: 0.004272
Epoch: 479	Training Loss: 0.008442	Validation Loss: 0.003692
Epoch: 480	Training Loss: 0.008440	Validation Loss: 0.006366
Epoch: 481	Training Loss: 0.008385	Validation Loss: 0.004655
Epoch: 482	Training Loss: 0.008365	Validation Loss: 0.005059
Epoch: 483	Training Loss: 0.008272	Validation Loss: 0.008517
Epoch: 484	Training Loss: 0.008411	Validation Loss: 0.003805
Epoch: 485	Training Loss: 0.008345	Validation Loss: 0.004926
Epoch: 486	Training Loss: 0.008466	Validation Loss: 0.003607
Epoch: 487	Training Loss: 0.008748	Validation Loss: 0.007465
Epoch: 488	Training Loss: 0.008289	Validation Loss: 0.005324
Epoch: 489	Training Loss: 0.008518	Validation Loss: 0.004304
Epoch: 490	Training Loss: 0.008343	Validation Loss: 0.005468
Epoch: 491	Training Loss: 0.008078	Validation Loss: 0.006590
Epoch: 492	Training Loss: 0.008359	Validation Loss: 0.005686
Epoch: 493	Training Loss: 0.008649	Validation Loss: 0.003506

Epoch: 494	Training Loss: 0.008870	Validation Loss: 0.006309
Epoch: 495	Training Loss: 0.008069	Validation Loss: 0.003375
Epoch: 496	Training Loss: 0.008056	Validation Loss: 0.006854
Epoch: 497	Training Loss: 0.008637	Validation Loss: 0.003165
Validation loss decreased (0.003246 --> 0.003165). Saving model ...		
Epoch: 498	Training Loss: 0.008124	Validation Loss: 0.005569
Epoch: 499	Training Loss: 0.008153	Validation Loss: 0.003614
Epoch: 500	Training Loss: 0.008401	Validation Loss: 0.005867

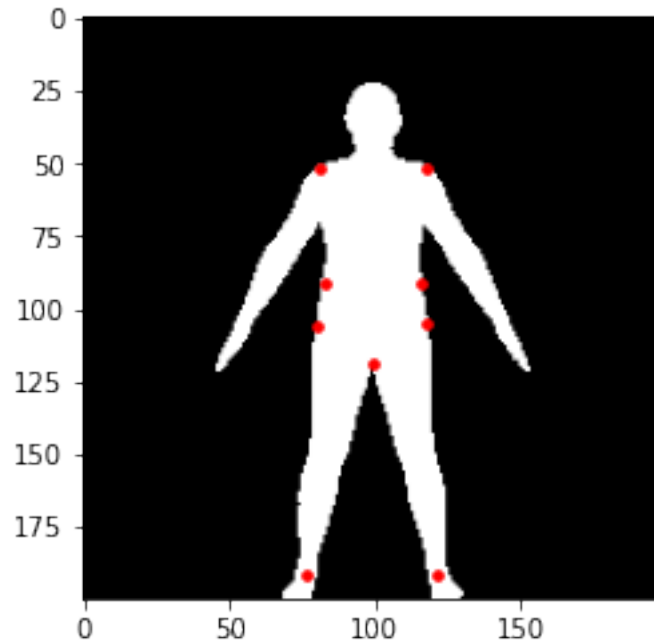
### 3.4 Check the model prediction ability

```
[0]: # Evaluate this one
model_front.load_state_dict(torch.load(ROOT_FOLDER+'model_front.pt'))
KPT_PREDICT = predict(test_loader_front, model_front)
```

```
[96]: # Draw the changing curve
n_epochs=500
x=range(0,n_epochs)
plt.figure(figsize=(12,8),dpi=80)
y1=train_losses
y2=valid_losses
plt.grid(alpha=0.4)
plt.plot(x,y1,label='train_losses',color='orange',linestyle=":",linewidth=2)
plt.plot(x,y2,label='valid_losses',color='cyan',linestyle="--",linewidth=3)
plt.xlabel('epoch',size=50)
plt.ylabel('losses',size=50)
plt.legend()
plt.show()
```



```
[97]: #Show one prediction  
idx = np.random.randint(KPT_PREDICT.shape[0])  
print(idx)  
draw_points(IMG_TEST[idx,:,:], KPT_PREDICT[idx,:].reshape((-1,2)))
```



## 4 Task 2: Image Generation

In the previous task, we predict low dimensional data (keypoints 9x2) from high dimensional input (images 200x200). Let's consider a different case now.

- Train a image generation model to predict silhouette images from keypoints.
- Input: 9x2 keypoints coordinates
- Output: corresponding silhouette images
- All of the available data can be used, including the keypoints you predicted on `imgs2.npy` in the previous task.
- Predict silhouette images for keypoints in `kpts3.npy`.
- Visualize the generated image together with input keypoints.

### 4.1 Define the training function for task 2

```
[0]: def train_img(train_loader, valid_loader, model, criterion, optimizer,
    n_epochs=50, saved_model='model.pt'):
    '''
    Train the model

    Args:
        train_loader (DataLoader): DataLoader for train Dataset
        valid_loader (DataLoader): DataLoader for valid Dataset
        model (nn.Module): model to be trained on
        criterion (torch.nn): loss function
```



```

    optimizer (torch.optim): optimization algorithms
    n_epochs (int): number of epochs to train the model
    saved_model (str): file path for saving model

Return:
    tuple of train_losses, valid_losses
'''

# initialize tracker for minimum validation loss
valid_loss_min = np.Inf # set initial "min" to infinity

train_losses = []
valid_losses = []

for epoch in range(n_epochs):
    # monitor training loss
    train_loss = 0.0
    valid_loss = 0.0

    #####
    # train the model #
    #####
    model.train() # prep model for training
    for batch in train_loader:
        # clear the gradients of all optimized variables
        optimizer.zero_grad()
        # forward pass: compute predicted outputs by passing inputs to the
→model
        output = model(batch['keypoints'].to(device))
        # calculate the loss
        img = batch['image']
        img = img.reshape([len(img),40000])
        loss = criterion(output, img.to(device))
        # backward pass: compute gradient of the loss with respect to model
→parameters
        loss.backward()
        # perform a single optimization step (parameter update)
        optimizer.step()
        # update running training loss
        train_loss += loss.item()*batch['keypoints'].size(0)

    #####
    # validate the model #
    #####
    model.eval() # prep model for evaluation
    for batch in valid_loader:

```

```

        # forward pass: compute predicted outputs by passing inputs to the
→model
        output = model(batch['keypoints'].to(device))
        # calculate the loss
        img2 = batch['image']
        img2 = img2.reshape([len(img2), 40000])
        loss = criterion(output, img2.to(device))
        # update running validation loss
        valid_loss += loss.item()*batch['keypoints'].size(0)

    # print training/validation statistics
    # calculate average Root Mean Square loss over an epoch
    train_loss = np.sqrt(train_loss/len(train_loader.sampler.indices))
    valid_loss = np.sqrt(valid_loss/len(valid_loader.sampler.indices))

    train_losses.append(train_loss)
    valid_losses.append(valid_loss)

    print('Epoch: {} \tTraining Loss: {:.6f} \tValidation Loss: {:.6f}'
          .format(epoch+1, train_loss, valid_loss))

    # save model if validation loss has decreased
    if valid_loss <= valid_loss_min:
        print('Validation loss decreased ({:.6f} --> {:.6f}). Saving model_
→...'
              .format(valid_loss_min, valid_loss))
        torch.save(model.state_dict(), saved_model)
        model_temp = model.state_dict()
        valid_loss_min = valid_loss
    return train_losses, valid_losses

```

## 4.2 Train the model using MLP

The following parameters are chosen:

- Learning rate: 0.001
- Number of epochs: 150
- Some pooling layers and convolutional layers in the CNN has been changed and can be found in the code.

```

[131]: device = torch.device('cuda' if torch.cuda.is_available() else 'CPU')
model_img = MLP(input_size=18, output_size=IMG_SIZE*IMG_SIZE,
                hidden_layers=[64, 128], drop_p=0.1)
model_img = model_img.to(device)
criterion = nn.MSELoss()
optimizer = optim.Adam(model_img.parameters(), lr=0.0001)
# optimizer = optim.SGD(model_img.parameters(), lr=0.001, momentum=0.95,
→weight_decay=1e-5)

```

```

train_losses, valid_losses = train_img(train_loader_front, valid_loader_front,
                                         model_img, criterion, optimizer,
                                         n_epochs=300,
                                         saved_model=ROOT_FOLDER+'model_adjusted.pt')

```

```

Epoch: 1      Training Loss: 0.382900      Validation Loss: 0.380007
Validation loss decreased (inf --> 0.380007). Saving model ...
Epoch: 2      Training Loss: 0.379832      Validation Loss: 0.377143
Validation loss decreased (0.380007 --> 0.377143). Saving model ...
Epoch: 3      Training Loss: 0.376955      Validation Loss: 0.373940
Validation loss decreased (0.377143 --> 0.373940). Saving model ...
Epoch: 4      Training Loss: 0.373745      Validation Loss: 0.370545
Validation loss decreased (0.373940 --> 0.370545). Saving model ...
Epoch: 5      Training Loss: 0.369799      Validation Loss: 0.366022
Validation loss decreased (0.370545 --> 0.366022). Saving model ...
Epoch: 6      Training Loss: 0.364614      Validation Loss: 0.360271
Validation loss decreased (0.366022 --> 0.360271). Saving model ...
Epoch: 7      Training Loss: 0.358361      Validation Loss: 0.353355
Validation loss decreased (0.360271 --> 0.353355). Saving model ...
Epoch: 8      Training Loss: 0.350934      Validation Loss: 0.344873
Validation loss decreased (0.353355 --> 0.344873). Saving model ...
Epoch: 9      Training Loss: 0.341577      Validation Loss: 0.334180
Validation loss decreased (0.344873 --> 0.334180). Saving model ...
Epoch: 10     Training Loss: 0.330707      Validation Loss: 0.322179
Validation loss decreased (0.334180 --> 0.322179). Saving model ...
Epoch: 11     Training Loss: 0.317634      Validation Loss: 0.307635
Validation loss decreased (0.322179 --> 0.307635). Saving model ...
Epoch: 12     Training Loss: 0.302924      Validation Loss: 0.291507
Validation loss decreased (0.307635 --> 0.291507). Saving model ...
Epoch: 13     Training Loss: 0.286546      Validation Loss: 0.273292
Validation loss decreased (0.291507 --> 0.273292). Saving model ...
Epoch: 14     Training Loss: 0.267945      Validation Loss: 0.254076
Validation loss decreased (0.273292 --> 0.254076). Saving model ...
Epoch: 15     Training Loss: 0.248308      Validation Loss: 0.233108
Validation loss decreased (0.254076 --> 0.233108). Saving model ...
Epoch: 16     Training Loss: 0.228959      Validation Loss: 0.213278
Validation loss decreased (0.233108 --> 0.213278). Saving model ...
Epoch: 17     Training Loss: 0.210488      Validation Loss: 0.193470
Validation loss decreased (0.213278 --> 0.193470). Saving model ...
Epoch: 18     Training Loss: 0.191765      Validation Loss: 0.175228
Validation loss decreased (0.193470 --> 0.175228). Saving model ...
Epoch: 19     Training Loss: 0.175993      Validation Loss: 0.160564
Validation loss decreased (0.175228 --> 0.160564). Saving model ...
Epoch: 20     Training Loss: 0.162689      Validation Loss: 0.148107
Validation loss decreased (0.160564 --> 0.148107). Saving model ...
Epoch: 21     Training Loss: 0.154043      Validation Loss: 0.138929

```

Validation loss decreased (0.148107 --> 0.138929). Saving model ...

Epoch: 22	Training Loss: 0.147381	Validation Loss: 0.131964
Validation loss decreased (0.138929 --> 0.131964). Saving model ...		
Epoch: 23	Training Loss: 0.143994	Validation Loss: 0.129216
Validation loss decreased (0.131964 --> 0.129216). Saving model ...		
Epoch: 24	Training Loss: 0.140671	Validation Loss: 0.126193
Validation loss decreased (0.129216 --> 0.126193). Saving model ...		
Epoch: 25	Training Loss: 0.137593	Validation Loss: 0.125156
Validation loss decreased (0.126193 --> 0.125156). Saving model ...		
Epoch: 26	Training Loss: 0.137084	Validation Loss: 0.122057
Validation loss decreased (0.125156 --> 0.122057). Saving model ...		
Epoch: 27	Training Loss: 0.136893	Validation Loss: 0.121060
Validation loss decreased (0.122057 --> 0.121060). Saving model ...		
Epoch: 28	Training Loss: 0.136311	Validation Loss: 0.121051
Validation loss decreased (0.121060 --> 0.121051). Saving model ...		
Epoch: 29	Training Loss: 0.135503	Validation Loss: 0.120152
Validation loss decreased (0.121051 --> 0.120152). Saving model ...		
Epoch: 30	Training Loss: 0.134952	Validation Loss: 0.120293
Epoch: 31	Training Loss: 0.134698	Validation Loss: 0.121233
Epoch: 32	Training Loss: 0.134474	Validation Loss: 0.120696
Epoch: 33	Training Loss: 0.133732	Validation Loss: 0.119223
Validation loss decreased (0.120152 --> 0.119223). Saving model ...		
Epoch: 34	Training Loss: 0.132940	Validation Loss: 0.118827
Validation loss decreased (0.119223 --> 0.118827). Saving model ...		
Epoch: 35	Training Loss: 0.133184	Validation Loss: 0.120945
Epoch: 36	Training Loss: 0.132702	Validation Loss: 0.119878
Epoch: 37	Training Loss: 0.133504	Validation Loss: 0.118692
Validation loss decreased (0.118827 --> 0.118692). Saving model ...		
Epoch: 38	Training Loss: 0.133372	Validation Loss: 0.119959
Epoch: 39	Training Loss: 0.133003	Validation Loss: 0.119662
Epoch: 40	Training Loss: 0.131728	Validation Loss: 0.119993
Epoch: 41	Training Loss: 0.132100	Validation Loss: 0.119941
Epoch: 42	Training Loss: 0.132326	Validation Loss: 0.120638
Epoch: 43	Training Loss: 0.131582	Validation Loss: 0.119773
Epoch: 44	Training Loss: 0.131456	Validation Loss: 0.118838
Epoch: 45	Training Loss: 0.132316	Validation Loss: 0.119901
Epoch: 46	Training Loss: 0.131521	Validation Loss: 0.119561
Epoch: 47	Training Loss: 0.131768	Validation Loss: 0.119539
Epoch: 48	Training Loss: 0.132149	Validation Loss: 0.119088
Epoch: 49	Training Loss: 0.131292	Validation Loss: 0.119815
Epoch: 50	Training Loss: 0.132354	Validation Loss: 0.119912
Epoch: 51	Training Loss: 0.131386	Validation Loss: 0.118957
Epoch: 52	Training Loss: 0.130564	Validation Loss: 0.120358
Epoch: 53	Training Loss: 0.131615	Validation Loss: 0.118932
Epoch: 54	Training Loss: 0.130477	Validation Loss: 0.120078
Epoch: 55	Training Loss: 0.130468	Validation Loss: 0.119124
Epoch: 56	Training Loss: 0.130216	Validation Loss: 0.119886
Epoch: 57	Training Loss: 0.130469	Validation Loss: 0.120180

Epoch: 58	Training Loss: 0.131278	Validation Loss: 0.119783
Epoch: 59	Training Loss: 0.129878	Validation Loss: 0.119653
Epoch: 60	Training Loss: 0.130193	Validation Loss: 0.118975
Epoch: 61	Training Loss: 0.130853	Validation Loss: 0.120113
Epoch: 62	Training Loss: 0.129794	Validation Loss: 0.119489
Epoch: 63	Training Loss: 0.130058	Validation Loss: 0.120220
Epoch: 64	Training Loss: 0.129854	Validation Loss: 0.118714
Epoch: 65	Training Loss: 0.129829	Validation Loss: 0.119067
Epoch: 66	Training Loss: 0.129345	Validation Loss: 0.119375
Epoch: 67	Training Loss: 0.129879	Validation Loss: 0.119420
Epoch: 68	Training Loss: 0.129132	Validation Loss: 0.120484
Epoch: 69	Training Loss: 0.129532	Validation Loss: 0.120108
Epoch: 70	Training Loss: 0.130402	Validation Loss: 0.120056
Epoch: 71	Training Loss: 0.129537	Validation Loss: 0.120287
Epoch: 72	Training Loss: 0.129467	Validation Loss: 0.119637
Epoch: 73	Training Loss: 0.128616	Validation Loss: 0.119583
Epoch: 74	Training Loss: 0.129032	Validation Loss: 0.117758
Validation loss decreased (0.118692 --> 0.117758). Saving model ...		
Epoch: 75	Training Loss: 0.128797	Validation Loss: 0.120492
Epoch: 76	Training Loss: 0.129117	Validation Loss: 0.120160
Epoch: 77	Training Loss: 0.128341	Validation Loss: 0.119970
Epoch: 78	Training Loss: 0.129408	Validation Loss: 0.120078
Epoch: 79	Training Loss: 0.128251	Validation Loss: 0.120081
Epoch: 80	Training Loss: 0.129367	Validation Loss: 0.119127
Epoch: 81	Training Loss: 0.128499	Validation Loss: 0.120208
Epoch: 82	Training Loss: 0.128454	Validation Loss: 0.119380
Epoch: 83	Training Loss: 0.128211	Validation Loss: 0.119384
Epoch: 84	Training Loss: 0.128842	Validation Loss: 0.119560
Epoch: 85	Training Loss: 0.128545	Validation Loss: 0.118294
Epoch: 86	Training Loss: 0.127296	Validation Loss: 0.119012
Epoch: 87	Training Loss: 0.128223	Validation Loss: 0.120146
Epoch: 88	Training Loss: 0.127944	Validation Loss: 0.119479
Epoch: 89	Training Loss: 0.127691	Validation Loss: 0.119817
Epoch: 90	Training Loss: 0.127161	Validation Loss: 0.118776
Epoch: 91	Training Loss: 0.128273	Validation Loss: 0.119240
Epoch: 92	Training Loss: 0.128200	Validation Loss: 0.120256
Epoch: 93	Training Loss: 0.127605	Validation Loss: 0.119143
Epoch: 94	Training Loss: 0.127465	Validation Loss: 0.120149
Epoch: 95	Training Loss: 0.127792	Validation Loss: 0.118097
Epoch: 96	Training Loss: 0.128109	Validation Loss: 0.120633
Epoch: 97	Training Loss: 0.127578	Validation Loss: 0.119954
Epoch: 98	Training Loss: 0.127367	Validation Loss: 0.118531
Epoch: 99	Training Loss: 0.127227	Validation Loss: 0.118223
Epoch: 100	Training Loss: 0.127694	Validation Loss: 0.119439
Epoch: 101	Training Loss: 0.127627	Validation Loss: 0.120615
Epoch: 102	Training Loss: 0.127391	Validation Loss: 0.118603
Epoch: 103	Training Loss: 0.127227	Validation Loss: 0.119092
Epoch: 104	Training Loss: 0.127485	Validation Loss: 0.119259

Epoch: 105	Training Loss: 0.127316	Validation Loss: 0.119099
Epoch: 106	Training Loss: 0.126806	Validation Loss: 0.119090
Epoch: 107	Training Loss: 0.126881	Validation Loss: 0.119926
Epoch: 108	Training Loss: 0.127141	Validation Loss: 0.119659
Epoch: 109	Training Loss: 0.126746	Validation Loss: 0.120026
Epoch: 110	Training Loss: 0.127096	Validation Loss: 0.119712
Epoch: 111	Training Loss: 0.126903	Validation Loss: 0.120574
Epoch: 112	Training Loss: 0.126931	Validation Loss: 0.117888
Epoch: 113	Training Loss: 0.126991	Validation Loss: 0.118696
Epoch: 114	Training Loss: 0.127006	Validation Loss: 0.119610
Epoch: 115	Training Loss: 0.127116	Validation Loss: 0.118506
Epoch: 116	Training Loss: 0.125829	Validation Loss: 0.119244
Epoch: 117	Training Loss: 0.126573	Validation Loss: 0.120369
Epoch: 118	Training Loss: 0.126085	Validation Loss: 0.119757
Epoch: 119	Training Loss: 0.126564	Validation Loss: 0.119460
Epoch: 120	Training Loss: 0.126388	Validation Loss: 0.118291
Epoch: 121	Training Loss: 0.126636	Validation Loss: 0.119777
Epoch: 122	Training Loss: 0.126262	Validation Loss: 0.120206
Epoch: 123	Training Loss: 0.125851	Validation Loss: 0.119669
Epoch: 124	Training Loss: 0.126306	Validation Loss: 0.118859
Epoch: 125	Training Loss: 0.125957	Validation Loss: 0.119215
Epoch: 126	Training Loss: 0.126325	Validation Loss: 0.119371
Epoch: 127	Training Loss: 0.126268	Validation Loss: 0.118998
Epoch: 128	Training Loss: 0.125496	Validation Loss: 0.119065
Epoch: 129	Training Loss: 0.125905	Validation Loss: 0.119708
Epoch: 130	Training Loss: 0.125039	Validation Loss: 0.119613
Epoch: 131	Training Loss: 0.125889	Validation Loss: 0.118578
Epoch: 132	Training Loss: 0.125493	Validation Loss: 0.120865
Epoch: 133	Training Loss: 0.126283	Validation Loss: 0.118452
Epoch: 134	Training Loss: 0.126423	Validation Loss: 0.119535
Epoch: 135	Training Loss: 0.126240	Validation Loss: 0.119471
Epoch: 136	Training Loss: 0.125399	Validation Loss: 0.119103
Epoch: 137	Training Loss: 0.125953	Validation Loss: 0.119233
Epoch: 138	Training Loss: 0.125253	Validation Loss: 0.119398
Epoch: 139	Training Loss: 0.125758	Validation Loss: 0.120082
Epoch: 140	Training Loss: 0.125280	Validation Loss: 0.119155
Epoch: 141	Training Loss: 0.126161	Validation Loss: 0.118451
Epoch: 142	Training Loss: 0.125256	Validation Loss: 0.119190
Epoch: 143	Training Loss: 0.124894	Validation Loss: 0.119716
Epoch: 144	Training Loss: 0.125616	Validation Loss: 0.119757
Epoch: 145	Training Loss: 0.125261	Validation Loss: 0.119958
Epoch: 146	Training Loss: 0.125543	Validation Loss: 0.118444
Epoch: 147	Training Loss: 0.125397	Validation Loss: 0.119557
Epoch: 148	Training Loss: 0.125035	Validation Loss: 0.119191
Epoch: 149	Training Loss: 0.125324	Validation Loss: 0.118291
Epoch: 150	Training Loss: 0.125514	Validation Loss: 0.119509
Epoch: 151	Training Loss: 0.125155	Validation Loss: 0.118396
Epoch: 152	Training Loss: 0.124967	Validation Loss: 0.118425

Epoch: 153	Training Loss: 0.124742	Validation Loss: 0.119771
Epoch: 154	Training Loss: 0.124744	Validation Loss: 0.118124
Epoch: 155	Training Loss: 0.125075	Validation Loss: 0.118662
Epoch: 156	Training Loss: 0.125376	Validation Loss: 0.118858
Epoch: 157	Training Loss: 0.124631	Validation Loss: 0.118252
Epoch: 158	Training Loss: 0.125747	Validation Loss: 0.119440
Epoch: 159	Training Loss: 0.125381	Validation Loss: 0.119265
Epoch: 160	Training Loss: 0.124930	Validation Loss: 0.119338
Epoch: 161	Training Loss: 0.124842	Validation Loss: 0.118445
Epoch: 162	Training Loss: 0.124621	Validation Loss: 0.119480
Epoch: 163	Training Loss: 0.124447	Validation Loss: 0.118337
Epoch: 164	Training Loss: 0.124767	Validation Loss: 0.118806
Epoch: 165	Training Loss: 0.124755	Validation Loss: 0.118443
Epoch: 166	Training Loss: 0.124082	Validation Loss: 0.117904
Epoch: 167	Training Loss: 0.124718	Validation Loss: 0.117687
Validation loss decreased (0.117758 --> 0.117687). Saving model ...		
Epoch: 168	Training Loss: 0.124772	Validation Loss: 0.119401
Epoch: 169	Training Loss: 0.124671	Validation Loss: 0.119221
Epoch: 170	Training Loss: 0.124996	Validation Loss: 0.118506
Epoch: 171	Training Loss: 0.123837	Validation Loss: 0.117755
Epoch: 172	Training Loss: 0.125162	Validation Loss: 0.119138
Epoch: 173	Training Loss: 0.124341	Validation Loss: 0.118755
Epoch: 174	Training Loss: 0.124597	Validation Loss: 0.119259
Epoch: 175	Training Loss: 0.124437	Validation Loss: 0.119862
Epoch: 176	Training Loss: 0.124071	Validation Loss: 0.118931
Epoch: 177	Training Loss: 0.124409	Validation Loss: 0.118920
Epoch: 178	Training Loss: 0.125189	Validation Loss: 0.118393
Epoch: 179	Training Loss: 0.124331	Validation Loss: 0.118773
Epoch: 180	Training Loss: 0.124712	Validation Loss: 0.120042
Epoch: 181	Training Loss: 0.124124	Validation Loss: 0.119378
Epoch: 182	Training Loss: 0.125008	Validation Loss: 0.119430
Epoch: 183	Training Loss: 0.124604	Validation Loss: 0.119021
Epoch: 184	Training Loss: 0.124452	Validation Loss: 0.118086
Epoch: 185	Training Loss: 0.124248	Validation Loss: 0.119290
Epoch: 186	Training Loss: 0.123918	Validation Loss: 0.117500
Validation loss decreased (0.117687 --> 0.117500). Saving model ...		
Epoch: 187	Training Loss: 0.124663	Validation Loss: 0.119553
Epoch: 188	Training Loss: 0.123471	Validation Loss: 0.117755
Epoch: 189	Training Loss: 0.124101	Validation Loss: 0.119940
Epoch: 190	Training Loss: 0.124150	Validation Loss: 0.118550
Epoch: 191	Training Loss: 0.123889	Validation Loss: 0.119131
Epoch: 192	Training Loss: 0.124194	Validation Loss: 0.117832
Epoch: 193	Training Loss: 0.124131	Validation Loss: 0.119064
Epoch: 194	Training Loss: 0.122769	Validation Loss: 0.118545
Epoch: 195	Training Loss: 0.123568	Validation Loss: 0.118851
Epoch: 196	Training Loss: 0.123514	Validation Loss: 0.118757
Epoch: 197	Training Loss: 0.123966	Validation Loss: 0.119480
Epoch: 198	Training Loss: 0.123493	Validation Loss: 0.119018

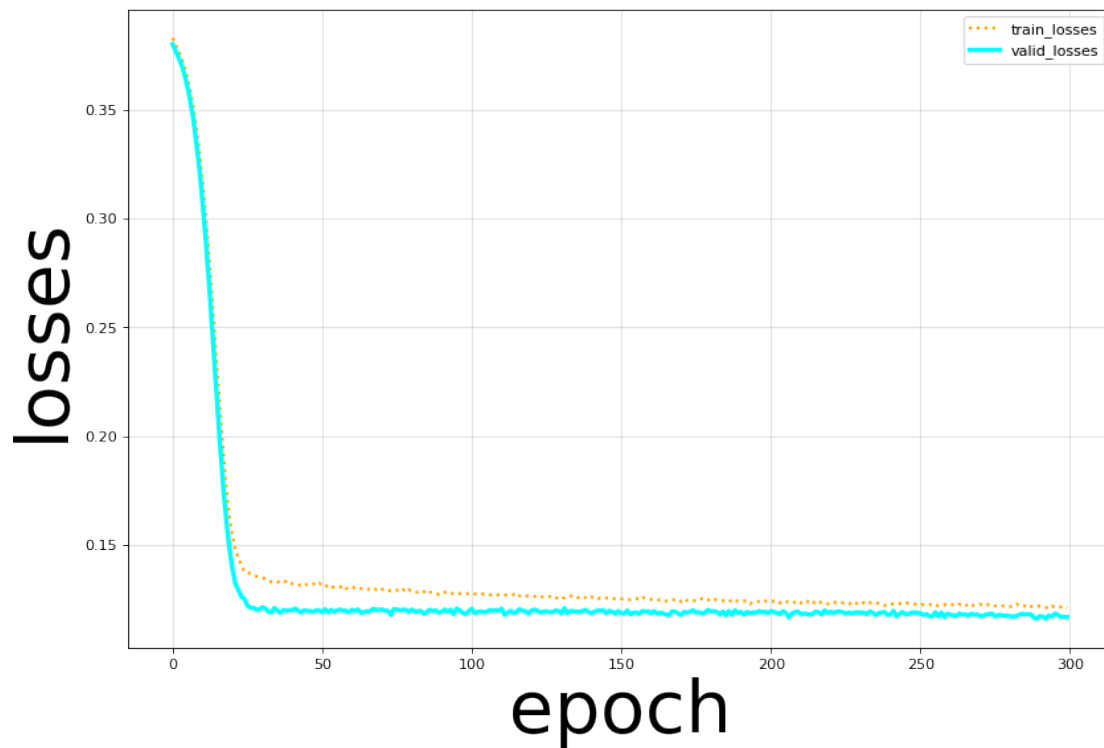
Epoch: 199	Training Loss: 0.124172	Validation Loss: 0.119212
Epoch: 200	Training Loss: 0.123702	Validation Loss: 0.118546
Epoch: 201	Training Loss: 0.123673	Validation Loss: 0.119460
Epoch: 202	Training Loss: 0.124171	Validation Loss: 0.119130
Epoch: 203	Training Loss: 0.124383	Validation Loss: 0.118741
Epoch: 204	Training Loss: 0.123249	Validation Loss: 0.118141
Epoch: 205	Training Loss: 0.123745	Validation Loss: 0.119422
Epoch: 206	Training Loss: 0.124236	Validation Loss: 0.118802
Epoch: 207	Training Loss: 0.123217	Validation Loss: 0.116537
Validation loss decreased (0.117500 --> 0.116537). Saving model ...		
Epoch: 208	Training Loss: 0.123189	Validation Loss: 0.118370
Epoch: 209	Training Loss: 0.123587	Validation Loss: 0.119012
Epoch: 210	Training Loss: 0.124093	Validation Loss: 0.119303
Epoch: 211	Training Loss: 0.123124	Validation Loss: 0.118241
Epoch: 212	Training Loss: 0.124490	Validation Loss: 0.118319
Epoch: 213	Training Loss: 0.124049	Validation Loss: 0.118243
Epoch: 214	Training Loss: 0.123689	Validation Loss: 0.117874
Epoch: 215	Training Loss: 0.122737	Validation Loss: 0.118254
Epoch: 216	Training Loss: 0.124348	Validation Loss: 0.119286
Epoch: 217	Training Loss: 0.123565	Validation Loss: 0.118866
Epoch: 218	Training Loss: 0.123532	Validation Loss: 0.118031
Epoch: 219	Training Loss: 0.123453	Validation Loss: 0.119790
Epoch: 220	Training Loss: 0.122756	Validation Loss: 0.118643
Epoch: 221	Training Loss: 0.123360	Validation Loss: 0.119208
Epoch: 222	Training Loss: 0.123143	Validation Loss: 0.118657
Epoch: 223	Training Loss: 0.123895	Validation Loss: 0.119173
Epoch: 224	Training Loss: 0.123414	Validation Loss: 0.119321
Epoch: 225	Training Loss: 0.123368	Validation Loss: 0.119028
Epoch: 226	Training Loss: 0.123590	Validation Loss: 0.118360
Epoch: 227	Training Loss: 0.122812	Validation Loss: 0.118877
Epoch: 228	Training Loss: 0.123221	Validation Loss: 0.118477
Epoch: 229	Training Loss: 0.123850	Validation Loss: 0.118793
Epoch: 230	Training Loss: 0.122510	Validation Loss: 0.118929
Epoch: 231	Training Loss: 0.123034	Validation Loss: 0.118796
Epoch: 232	Training Loss: 0.123368	Validation Loss: 0.118788
Epoch: 233	Training Loss: 0.122613	Validation Loss: 0.118107
Epoch: 234	Training Loss: 0.123413	Validation Loss: 0.119286
Epoch: 235	Training Loss: 0.122789	Validation Loss: 0.118866
Epoch: 236	Training Loss: 0.123248	Validation Loss: 0.117658
Epoch: 237	Training Loss: 0.123100	Validation Loss: 0.118544
Epoch: 238	Training Loss: 0.122841	Validation Loss: 0.118820
Epoch: 239	Training Loss: 0.122066	Validation Loss: 0.119091
Epoch: 240	Training Loss: 0.122608	Validation Loss: 0.117714
Epoch: 241	Training Loss: 0.122414	Validation Loss: 0.117880
Epoch: 242	Training Loss: 0.122958	Validation Loss: 0.119671
Epoch: 243	Training Loss: 0.122732	Validation Loss: 0.117091
Epoch: 244	Training Loss: 0.122873	Validation Loss: 0.118236
Epoch: 245	Training Loss: 0.123574	Validation Loss: 0.119080



Epoch: 246	Training Loss: 0.123467	Validation Loss: 0.118128
Epoch: 247	Training Loss: 0.123140	Validation Loss: 0.118125
Epoch: 248	Training Loss: 0.122817	Validation Loss: 0.118785
Epoch: 249	Training Loss: 0.122600	Validation Loss: 0.118842
Epoch: 250	Training Loss: 0.122973	Validation Loss: 0.118768
Epoch: 251	Training Loss: 0.122731	Validation Loss: 0.118488
Epoch: 252	Training Loss: 0.122495	Validation Loss: 0.117651
Epoch: 253	Training Loss: 0.122656	Validation Loss: 0.118100
Epoch: 254	Training Loss: 0.122618	Validation Loss: 0.117113
Epoch: 255	Training Loss: 0.122192	Validation Loss: 0.117118
Epoch: 256	Training Loss: 0.122655	Validation Loss: 0.118947
Epoch: 257	Training Loss: 0.121815	Validation Loss: 0.117064
Epoch: 258	Training Loss: 0.121990	Validation Loss: 0.118864
Epoch: 259	Training Loss: 0.122406	Validation Loss: 0.116700
Epoch: 260	Training Loss: 0.122914	Validation Loss: 0.118318
Epoch: 261	Training Loss: 0.122817	Validation Loss: 0.117846
Epoch: 262	Training Loss: 0.122279	Validation Loss: 0.117571
Epoch: 263	Training Loss: 0.121995	Validation Loss: 0.117242
Epoch: 264	Training Loss: 0.122695	Validation Loss: 0.118415
Epoch: 265	Training Loss: 0.122391	Validation Loss: 0.117914
Epoch: 266	Training Loss: 0.122877	Validation Loss: 0.117999
Epoch: 267	Training Loss: 0.121338	Validation Loss: 0.117463
Epoch: 268	Training Loss: 0.122098	Validation Loss: 0.118254
Epoch: 269	Training Loss: 0.122181	Validation Loss: 0.117858
Epoch: 270	Training Loss: 0.122441	Validation Loss: 0.117886
Epoch: 271	Training Loss: 0.122699	Validation Loss: 0.117281
Epoch: 272	Training Loss: 0.121989	Validation Loss: 0.117278
Epoch: 273	Training Loss: 0.122020	Validation Loss: 0.117961
Epoch: 274	Training Loss: 0.122664	Validation Loss: 0.118140
Epoch: 275	Training Loss: 0.122554	Validation Loss: 0.118324
Epoch: 276	Training Loss: 0.121847	Validation Loss: 0.117051
Epoch: 277	Training Loss: 0.122092	Validation Loss: 0.117655
Epoch: 278	Training Loss: 0.122017	Validation Loss: 0.117718
Epoch: 279	Training Loss: 0.121819	Validation Loss: 0.117322
Epoch: 280	Training Loss: 0.121500	Validation Loss: 0.117399
Epoch: 281	Training Loss: 0.121600	Validation Loss: 0.117129
Epoch: 282	Training Loss: 0.121668	Validation Loss: 0.116979
Epoch: 283	Training Loss: 0.122836	Validation Loss: 0.117193
Epoch: 284	Training Loss: 0.122140	Validation Loss: 0.117760
Epoch: 285	Training Loss: 0.121688	Validation Loss: 0.118248
Epoch: 286	Training Loss: 0.121837	Validation Loss: 0.117994
Epoch: 287	Training Loss: 0.120863	Validation Loss: 0.118089
Epoch: 288	Training Loss: 0.121662	Validation Loss: 0.117993
Epoch: 289	Training Loss: 0.121394	Validation Loss: 0.116440
Validation loss decreased (0.116537 --> 0.116440). Saving model ...		
Epoch: 290	Training Loss: 0.121850	Validation Loss: 0.116105
Validation loss decreased (0.116440 --> 0.116105). Saving model ...		
Epoch: 291	Training Loss: 0.121780	Validation Loss: 0.117189

Epoch: 292	Training Loss: 0.121718	Validation Loss: 0.116901
Epoch: 293	Training Loss: 0.121159	Validation Loss: 0.116041
Validation loss decreased (0.116105 --> 0.116041). Saving model ...		
Epoch: 294	Training Loss: 0.120647	Validation Loss: 0.117505
Epoch: 295	Training Loss: 0.121803	Validation Loss: 0.116976
Epoch: 296	Training Loss: 0.121221	Validation Loss: 0.118390
Epoch: 297	Training Loss: 0.121248	Validation Loss: 0.117742
Epoch: 298	Training Loss: 0.121388	Validation Loss: 0.116746
Epoch: 299	Training Loss: 0.121738	Validation Loss: 0.116857
Epoch: 300	Training Loss: 0.120874	Validation Loss: 0.116780

```
[132]: # Draw the changing curve
n_epochs=300
x=range(0,n_epochs)
plt.figure(figsize=(12,8),dpi=80)
y1=train_losses
y2=valid_losses
plt.grid(alpha=0.4)
plt.plot(x,y1,label='train_losses',color='orange',linestyle=":",linewidth=2)
plt.plot(x,y2,label='valid_losses',color='cyan',linestyle="-",linewidth=3)
plt.xlabel('epoch',size=50)
plt.ylabel('losses',size=50)
plt.legend()
plt.show()
```



### 4.3 Define the prediction function and test dataloader for task 2

```
[0]: def predict_img(data_loader, model):  
    '''  
    Predict keypoints  
    Args:  
        data_loader (DataLoader): DataLoader for Dataset  
        model (nn.Module): trained model for prediction.  
    Return:  
        predictions (array-like): keypoints in float (no. of images x  
→keypoints).  
    '''  
  
    model.eval() # prep model for evaluation  
  
    with torch.no_grad():  
        for i, batch in enumerate(data_loader):  
            # forward pass: compute predicted outputs by passing inputs to the  
→model  
            output = model(batch.to(device)).cpu().numpy()  
            if i == 0:  
                predictions = output  
            else:  
                predictions = np.vstack((predictions, output))  
  
    return predictions
```

```
[0]: class KeypointsDataset_kpt(Dataset):  
    '''Keypoints Dataset'''  
    def __init__(self, kpt, train=True, transform=None):  
        self.kpt = kpt  
        self.train = train  
        self.transform = transform  
  
    def __len__(self):  
        return self.kpt.shape[0]  
  
    def __getitem__(self, idx):  
        keypoints = self.kpt[idx,:,:].ravel().astype(np.float32)  
        sample = keypoints  
        if self.transform:  
            sample = self.transform(sample)  
        return sample  
  
class ToTensor(object):
```

```

'''Convert ndarrays in sample to Tensors.'''
def __call__(self, sample):
    if sample is not None:
        keypoints = torch.from_numpy(sample)
        return keypoints
    else:
        return

```

#### 4.4 Load test data and predict the images for given keypoints

```

[150]: print(KPT_TEST.shape)
        print(KPT_PREDICT.shape)
        a=KPT_PREDICT.reshape((-1,9,2))
        print(a.shape)

```

```

(100, 9, 2)
(1000, 18)
(1000, 9, 2)

```

```

[0]: tsfm_test_kpt = transforms.Compose([ToTensor()])

# Add the prediction result of the keypoints based on the imgs2
np.append(a,KPT_TEST)

testset_img = KeypointsDataset_kpt(KPT_TEST, train=False,
    ↪transform=tsfm_test_kpt)
test_loader_img = torch.utils.data.DataLoader(testset_img,
    ↪batch_size=batch_size)

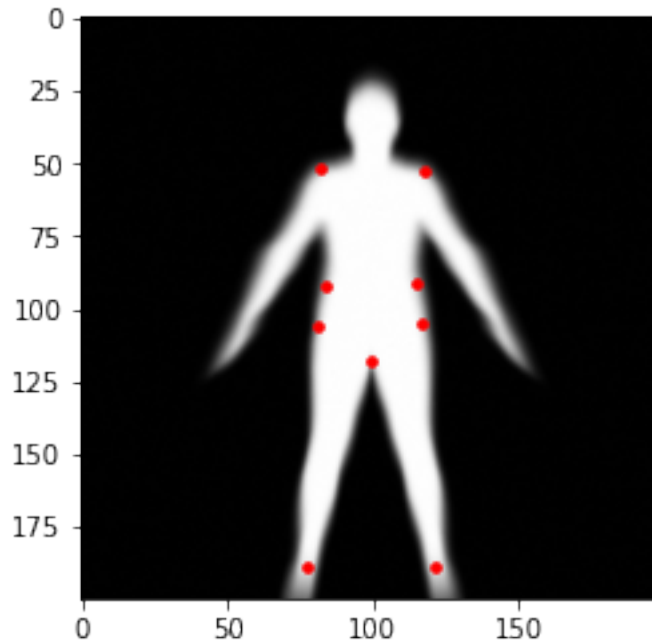
model_img.load_state_dict(torch.load(ROOT_FOLDER+'model_adjusted.pt'))
IMG_PREDICT = predict_img(test_loader_img, model_img)

```

```

[156]: #Show one prediction
        idx = np.random.randint(IMG_PREDICT.shape[0])
        print(idx)
        draw_points(IMG_PREDICT[idx,:].reshape(-1,200), KPT_TRAIN[idx,:,:])

```



## 5 Conclusion

During the assignment 4, I have learned a lot about the deep learning basic concepts including CNN structure, MLP structure. Now based on the assignment2, using the MLP method I have finished the image generation function even it is not a good solution to finish it. Maybe the random-forest is the best way to set up the prediction model with feature engineering based on assignment3 but we just have a week left, it is fine to finish the task in such a way.

I have learned a lot in the MAEG5735 from Prof. Wang and TA. Liu, here to say thanks sincerely to them.

[162]: 

```
!apt-get install texlive texlive-xetex texlive-latex-extra pandoc
!pip install pypandoc
```

Reading package lists... Done

Building dependency tree

Reading state information... Done

pandoc is already the newest version (1.19.2.4~dfsg-1build4).

pandoc set to manually installed.

The following additional packages will be installed:

```
fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre
javascript-common libcups2 libcupsfilters1 libcupsimage2 libgs9
libgs9-common libijs-0.35 libjbig2dec0 libjs-jquery libkpathsea6 libpotrace0
libptexenc1 libruby2.5 libsynchronet1 libtexlua52 libtexluaajit2 libzip-0-13
lmodern poppler-data preview-latex-style rake ruby ruby-did-you-mean
ruby-minitest ruby-net-telnet ruby-power-assert ruby-test-unit ruby2.5
```

rubygems-integration tlutils tex-common tex-gyre texlive-base  
texlive-binaries texlive-fonts-recommended texlive-latex-base  
texlive-latex-recommended texlive-pictures texlive-plain-generic tipa

Suggested packages:

fonts-noto apache2 | lighttpd | httpd cups-common poppler-utils ghostscript  
fonts-japanese-mincho | fonts-ipafont-mincho fonts-japanese-gothic  
| fonts-ipafont-gothic fonts-arphic-ukai fonts-arphic-uming fonts-nanum ri  
ruby-dev bundler debhelper gv | postscript-viewer perl-tk xpdf-reader  
| pdf-viewer texlive-fonts-recommended-doc texlive-latex-base-doc  
python-pygments icc-profiles libfile-which-perl  
libspreadsheet-parseexcel-perl texlive-latex-extra-doc  
texlive-latex-recommended-doc texlive-pstricks dot2tex prerex ruby-tcltk  
| libtcltk-ruby texlive-pictures-doc vprerex

The following NEW packages will be installed:

fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre  
javascript-common libcupsfilters1 libcupsimage2 libgs9 libgs9-common  
libijs-0.35 libjbig2dec0 libjs-jquery libkpathsea6 libpotrace0 libptexenc1  
libruby2.5 libsynchronet1 libtexlua52 libtexluajit2 libzzip-0-13 lmodern  
poppler-data preview-latex-style rake ruby ruby-did-you-mean ruby-minitest  
ruby-net-telnet ruby-power-assert ruby-test-unit ruby2.5  
rubygems-integration tlutils tex-common tex-gyre texlive texlive-base  
texlive-binaries texlive-fonts-recommended texlive-latex-base  
texlive-latex-extra texlive-latex-recommended texlive-pictures  
texlive-plain-generic texlive-xetex tipa

The following packages will be upgraded:

libcups2

1 upgraded, 47 newly installed, 0 to remove and 107 not upgraded.

Need to get 146 MB of archives.

After this operation, 460 MB of additional disk space will be used.

Get:1 <http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-droid-fallback>  
all 1:6.0.1r16-1.1 [1,805 kB]

Get:2 <http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-lato> all 2.0-2  
[2,698 kB]

Get:3 <http://archive.ubuntu.com/ubuntu bionic/main amd64 poppler-data> all  
0.4.8-2 [1,479 kB]

Get:4 <http://archive.ubuntu.com/ubuntu bionic/main amd64 tex-common> all 6.09  
[33.0 kB]

Get:5 <http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-lmodern> all  
2.004.5-3 [4,551 kB]

Get:6 <http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-noto-mono> all  
20171026-2 [75.5 kB]

Get:7 <http://archive.ubuntu.com/ubuntu bionic/universe amd64 fonts-texgyre> all  
20160520-1 [8,761 kB]

Get:8 <http://archive.ubuntu.com/ubuntu bionic/main amd64 javascript-common> all  
11 [6,066 B]

Get:9 <http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libcups2> amd64  
2.2.7-1ubuntu2.8 [211 kB]

Get:10 <http://archive.ubuntu.com/ubuntu bionic-updates/main amd64>

libcupsfilters1 amd64 1.20.2-0ubuntu3.1 [108 kB]  
 Get:11 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libcupsimage2  
 amd64 2.2.7-1ubuntu2.8 [18.6 kB]  
 Get:12 http://archive.ubuntu.com/ubuntu bionic/main amd64 libijs-0.35 amd64  
 0.35-13 [15.5 kB]  
 Get:13 http://archive.ubuntu.com/ubuntu bionic/main amd64 libjbig2dec0 amd64  
 0.13-6 [55.9 kB]  
 Get:14 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libgs9-common  
 all 9.26~dfsg+0-0ubuntu0.18.04.12 [5,092 kB]  
 Get:15 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libgs9 amd64  
 9.26~dfsg+0-0ubuntu0.18.04.12 [2,264 kB]  
 Get:16 http://archive.ubuntu.com/ubuntu bionic/main amd64 libjs-jquery all  
 3.2.1-1 [152 kB]  
 Get:17 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libkpathsea6  
 amd64 2017.20170613.44572-8ubuntu0.1 [54.9 kB]  
 Get:18 http://archive.ubuntu.com/ubuntu bionic/main amd64 libpotrace0 amd64  
 1.14-2 [17.4 kB]  
 Get:19 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libptexenc1  
 amd64 2017.20170613.44572-8ubuntu0.1 [34.5 kB]  
 Get:20 http://archive.ubuntu.com/ubuntu bionic/main amd64 rubygems-integration  
 all 1.11 [4,994 B]  
 Get:21 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 ruby2.5 amd64  
 2.5.1-1ubuntu1.6 [48.6 kB]  
 Get:22 http://archive.ubuntu.com/ubuntu bionic/main amd64 ruby amd64 1:2.5.1  
 [5,712 B]  
 Get:23 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 rake all  
 12.3.1-1ubuntu0.1 [44.9 kB]  
 Get:24 http://archive.ubuntu.com/ubuntu bionic/main amd64 ruby-did-you-mean all  
 1.2.0-2 [9,700 B]  
 Get:25 http://archive.ubuntu.com/ubuntu bionic/main amd64 ruby-minitest all  
 5.10.3-1 [38.6 kB]  
 Get:26 http://archive.ubuntu.com/ubuntu bionic/main amd64 ruby-net-telnet all  
 0.1.1-2 [12.6 kB]  
 Get:27 http://archive.ubuntu.com/ubuntu bionic/main amd64 ruby-power-assert all  
 0.3.0-1 [7,952 B]  
 Get:28 http://archive.ubuntu.com/ubuntu bionic/main amd64 ruby-test-unit all  
 3.2.5-1 [61.1 kB]  
 Get:29 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libruby2.5  
 amd64 2.5.1-1ubuntu1.6 [3,069 kB]  
 Get:30 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libsyntax1  
 amd64 2017.20170613.44572-8ubuntu0.1 [41.4 kB]  
 Get:31 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libtexlua52  
 amd64 2017.20170613.44572-8ubuntu0.1 [91.2 kB]  
 Get:32 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libtexluajit2  
 amd64 2017.20170613.44572-8ubuntu0.1 [230 kB]  
 Get:33 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libzip-0-13  
 amd64 0.13.62-3.1ubuntu0.18.04.1 [26.0 kB]  
 Get:34 http://archive.ubuntu.com/ubuntu bionic/main amd64 lmodern all 2.004.5-3

```

[9,631 kB]
Get:35 http://archive.ubuntu.com/ubuntu bionic/main amd64 preview-latex-style
all 11.91-1ubuntu1 [185 kB]
Get:36 http://archive.ubuntu.com/ubuntu bionic/main amd64 t1utils amd64 1.41-2
[56.0 kB]
Get:37 http://archive.ubuntu.com/ubuntu bionic/universe amd64 tex-gyre all
20160520-1 [4,998 kB]
Get:38 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 texlive-
binaries amd64 2017.20170613.44572-8ubuntu0.1 [8,179 kB]
Get:39 http://archive.ubuntu.com/ubuntu bionic/main amd64 texlive-base all
2017.20180305-1 [18.7 MB]
Get:40 http://archive.ubuntu.com/ubuntu bionic/universe amd64 texlive-fonts-
recommended all 2017.20180305-1 [5,262 kB]
Get:41 http://archive.ubuntu.com/ubuntu bionic/main amd64 texlive-latex-base all
2017.20180305-1 [951 kB]
Get:42 http://archive.ubuntu.com/ubuntu bionic/main amd64 texlive-latex-
recommended all 2017.20180305-1 [14.9 MB]
Get:43 http://archive.ubuntu.com/ubuntu bionic/universe amd64 texlive all
2017.20180305-1 [14.4 kB]
Get:44 http://archive.ubuntu.com/ubuntu bionic/universe amd64 texlive-pictures
all 2017.20180305-1 [4,026 kB]
Get:45 http://archive.ubuntu.com/ubuntu bionic/universe amd64 texlive-latex-
extra all 2017.20180305-2 [10.6 MB]
Get:46 http://archive.ubuntu.com/ubuntu bionic/universe amd64 texlive-plain-
generic all 2017.20180305-2 [23.6 MB]
Get:47 http://archive.ubuntu.com/ubuntu bionic/universe amd64 tipa all 2:1.3-20
[2,978 kB]
Get:48 http://archive.ubuntu.com/ubuntu bionic/universe amd64 texlive-xetex all
2017.20180305-1 [10.7 MB]
Fetched 146 MB in 16s (9,004 kB/s)
Extracting templates from packages: 100%
Preconfiguring packages ...
Selecting previously unselected package fonts-droid-fallback.
(Reading database ... 144568 files and directories currently installed.)
Preparing to unpack .../00-fonts-droid-fallback_1%3a6.0.1r16-1.1_all.deb ...
Unpacking fonts-droid-fallback (1:6.0.1r16-1.1) ...
Selecting previously unselected package fonts-lato.
Preparing to unpack .../01-fonts-lato_2.0-2_all.deb ...
Unpacking fonts-lato (2.0-2) ...
Selecting previously unselected package poppler-data.
Preparing to unpack .../02-poppler-data_0.4.8-2_all.deb ...
Unpacking poppler-data (0.4.8-2) ...
Selecting previously unselected package tex-common.
Preparing to unpack .../03-tex-common_6.09_all.deb ...
Unpacking tex-common (6.09) ...
Selecting previously unselected package fonts-lmodern.
Preparing to unpack .../04-fonts-lmodern_2.004.5-3_all.deb ...
Unpacking fonts-lmodern (2.004.5-3) ...

```



```

Selecting previously unselected package fonts-noto-mono.
Preparing to unpack .../05-fonts-noto-mono_20171026-2_all.deb ...
Unpacking fonts-noto-mono (20171026-2) ...
Selecting previously unselected package fonts-texgyre.
Preparing to unpack .../06-fonts-texgyre_20160520-1_all.deb ...
Unpacking fonts-texgyre (20160520-1) ...
Selecting previously unselected package javascript-common.
Preparing to unpack .../07-javascript-common_11_all.deb ...
Unpacking javascript-common (11) ...
Preparing to unpack .../08-libcups2_2.2.7-1ubuntu2.8_amd64.deb ...
Unpacking libcups2:amd64 (2.2.7-1ubuntu2.8) over (2.2.7-1ubuntu2.7) ...
Selecting previously unselected package libcupsfilters1:amd64.
Preparing to unpack .../09-libcupsfilters1_1.20.2-0ubuntu3.1_amd64.deb ...
Unpacking libcupsfilters1:amd64 (1.20.2-0ubuntu3.1) ...
Selecting previously unselected package libcupsimage2:amd64.
Preparing to unpack .../10-libcupsimage2_2.2.7-1ubuntu2.8_amd64.deb ...
Unpacking libcupsimage2:amd64 (2.2.7-1ubuntu2.8) ...
Selecting previously unselected package libijs-0.35:amd64.
Preparing to unpack .../11-libijs-0.35_0.35-13_amd64.deb ...
Unpacking libijs-0.35:amd64 (0.35-13) ...
Selecting previously unselected package libjbig2dec0:amd64.
Preparing to unpack .../12-libjbig2dec0_0.13-6_amd64.deb ...
Unpacking libjbig2dec0:amd64 (0.13-6) ...
Selecting previously unselected package libgs9-common.
Preparing to unpack .../13-libgs9-common_9.26~dfsg+0-0ubuntu0.18.04.12_all.deb
...
Unpacking libgs9-common (9.26~dfsg+0-0ubuntu0.18.04.12) ...
Selecting previously unselected package libgs9:amd64.
Preparing to unpack .../14-libgs9_9.26~dfsg+0-0ubuntu0.18.04.12_amd64.deb ...
Unpacking libgs9:amd64 (9.26~dfsg+0-0ubuntu0.18.04.12) ...
Selecting previously unselected package libjs-jquery.
Preparing to unpack .../15-libjs-jquery_3.2.1-1_all.deb ...
Unpacking libjs-jquery (3.2.1-1) ...
Selecting previously unselected package libkpathsea6:amd64.
Preparing to unpack .../16-libkpathsea6_2017.20170613.44572-8ubuntu0.1_amd64.deb
...
Unpacking libkpathsea6:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Selecting previously unselected package libpotrace0.
Preparing to unpack .../17-libpotrace0_1.14-2_amd64.deb ...
Unpacking libpotrace0 (1.14-2) ...
Selecting previously unselected package libptexenc1:amd64.
Preparing to unpack .../18-libptexenc1_2017.20170613.44572-8ubuntu0.1_amd64.deb
...
Unpacking libptexenc1:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Selecting previously unselected package rubygems-integration.
Preparing to unpack .../19-rubygems-integration_1.11_all.deb ...
Unpacking rubygems-integration (1.11) ...
Selecting previously unselected package ruby2.5.

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Preparing to unpack .../20-ruby2.5_2.5.1-1ubuntu1.6_amd64.deb ...
Unpacking ruby2.5 (2.5.1-1ubuntu1.6) ...
Selecting previously unselected package ruby.
Preparing to unpack .../21-ruby_1%3a2.5.1_amd64.deb ...
Unpacking ruby (1:2.5.1) ...
Selecting previously unselected package rake.
Preparing to unpack .../22-rake_12.3.1-1ubuntu0.1_all.deb ...
Unpacking rake (12.3.1-1ubuntu0.1) ...
Selecting previously unselected package ruby-did-you-mean.
Preparing to unpack .../23-ruby-did-you-mean_1.2.0-2_all.deb ...
Unpacking ruby-did-you-mean (1.2.0-2) ...
Selecting previously unselected package ruby-minitest.
Preparing to unpack .../24-ruby-minitest_5.10.3-1_all.deb ...
Unpacking ruby-minitest (5.10.3-1) ...
Selecting previously unselected package ruby-net-telnet.
Preparing to unpack .../25-ruby-net-telnet_0.1.1-2_all.deb ...
Unpacking ruby-net-telnet (0.1.1-2) ...
Selecting previously unselected package ruby-power-assert.
Preparing to unpack .../26-ruby-power-assert_0.3.0-1_all.deb ...
Unpacking ruby-power-assert (0.3.0-1) ...
Selecting previously unselected package ruby-test-unit.
Preparing to unpack .../27-ruby-test-unit_3.2.5-1_all.deb ...
Unpacking ruby-test-unit (3.2.5-1) ...
Selecting previously unselected package libruby2.5:amd64.
Preparing to unpack .../28-libruby2.5_2.5.1-1ubuntu1.6_amd64.deb ...
Unpacking libruby2.5:amd64 (2.5.1-1ubuntu1.6) ...
Selecting previously unselected package libsyntax1:amd64.
Preparing to unpack .../29-libsyntax1_2017.20170613.44572-8ubuntu0.1_amd64.deb
...
Unpacking libsyntax1:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Selecting previously unselected package libtexlua52:amd64.
Preparing to unpack .../30-libtexlua52_2017.20170613.44572-8ubuntu0.1_amd64.deb
...
Unpacking libtexlua52:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Selecting previously unselected package libtexluajit2:amd64.
Preparing to unpack
.../31-libtexluajit2_2017.20170613.44572-8ubuntu0.1_amd64.deb ...
Unpacking libtexluajit2:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Selecting previously unselected package libzip-0-13:amd64.
Preparing to unpack .../32-libzip-0-13_0.13.62-3.1ubuntu0.18.04.1_amd64.deb ...
Unpacking libzip-0-13:amd64 (0.13.62-3.1ubuntu0.18.04.1) ...
Selecting previously unselected package lmodern.
Preparing to unpack .../33-lmodern_2.004.5-3_all.deb ...
Unpacking lmodern (2.004.5-3) ...
Selecting previously unselected package preview-latex-style.
Preparing to unpack .../34-preview-latex-style_11.91-1ubuntu1_all.deb ...
Unpacking preview-latex-style (11.91-1ubuntu1) ...
Selecting previously unselected package tiutils.

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Preparing to unpack .../35-tlutils_1.41-2_amd64.deb ...
Unpacking tlutils (1.41-2) ...
Selecting previously unselected package tex-gyre.
Preparing to unpack .../36-tex-gyre_20160520-1_all.deb ...
Unpacking tex-gyre (20160520-1) ...
Selecting previously unselected package texlive-binaries.
Preparing to unpack .../37-texlive-
binaries_2017.20170613.44572-8ubuntu0.1_amd64.deb ...
Unpacking texlive-binaries (2017.20170613.44572-8ubuntu0.1) ...
Selecting previously unselected package texlive-base.
Preparing to unpack .../38-texlive-base_2017.20180305-1_all.deb ...
Unpacking texlive-base (2017.20180305-1) ...
Selecting previously unselected package texlive-fonts-recommended.
Preparing to unpack .../39-texlive-fonts-recommended_2017.20180305-1_all.deb ...
Unpacking texlive-fonts-recommended (2017.20180305-1) ...
Selecting previously unselected package texlive-latex-base.
Preparing to unpack .../40-texlive-latex-base_2017.20180305-1_all.deb ...
Unpacking texlive-latex-base (2017.20180305-1) ...
Selecting previously unselected package texlive-latex-recommended.
Preparing to unpack .../41-texlive-latex-recommended_2017.20180305-1_all.deb ...
Unpacking texlive-latex-recommended (2017.20180305-1) ...
Selecting previously unselected package texlive.
Preparing to unpack .../42-texlive_2017.20180305-1_all.deb ...
Unpacking texlive (2017.20180305-1) ...
Selecting previously unselected package texlive-pictures.
Preparing to unpack .../43-texlive-pictures_2017.20180305-1_all.deb ...
Unpacking texlive-pictures (2017.20180305-1) ...
Selecting previously unselected package texlive-latex-extra.
Preparing to unpack .../44-texlive-latex-extra_2017.20180305-2_all.deb ...
Unpacking texlive-latex-extra (2017.20180305-2) ...
Selecting previously unselected package texlive-plain-generic.
Preparing to unpack .../45-texlive-plain-generic_2017.20180305-2_all.deb ...
Unpacking texlive-plain-generic (2017.20180305-2) ...
Selecting previously unselected package tipa.
Preparing to unpack .../46-tipa_2%3a1.3-20_all.deb ...
Unpacking tipa (2:1.3-20) ...
Selecting previously unselected package texlive-xetex.
Preparing to unpack .../47-texlive-xetex_2017.20180305-1_all.deb ...
Unpacking texlive-xetex (2017.20180305-1) ...
Setting up libgs9-common (9.26~dfsg+0-0ubuntu0.18.04.12) ...
Setting up libkpathsea6:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Setting up libjs-jquery (3.2.1-1) ...
Setting up libtexlua52:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Setting up fonts-droid-fallback (1:6.0.1r16-1.1) ...
Setting up libsynctex1:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Setting up libptexenc1:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Setting up tex-common (6.09) ...
update-language: texlive-base not installed and configured, doing nothing!

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Setting up poppler-data (0.4.8-2) ...
Setting up tex-gyre (20160520-1) ...
Setting up preview-latex-style (11.91-1ubuntu1) ...
Setting up fonts-texgyre (20160520-1) ...
Setting up fonts-noto-mono (20171026-2) ...
Setting up fonts-lato (2.0-2) ...
Setting up libjbig2dec0:amd64 (0.13-6) ...
Setting up ruby-did-you-mean (1.2.0-2) ...
Setting up tlutils (1.41-2) ...
Setting up ruby-net-telnet (0.1.1-2) ...
Setting up libijs-0.35:amd64 (0.35-13) ...
Setting up rubygems-integration (1.11) ...
Setting up libpotrace0 (1.14-2) ...
Setting up javascript-common (11) ...
Setting up ruby-minitest (5.10.3-1) ...
Setting up libzip-0-13:amd64 (0.13.62-3.1ubuntu0.18.04.1) ...
Setting up libtexluaajit2:amd64 (2017.20170613.44572-8ubuntu0.1) ...
Setting up fonts-lmodern (2.004.5-3) ...
Setting up ruby-power-assert (0.3.0-1) ...
Setting up libcups2:amd64 (2.2.7-1ubuntu2.8) ...
Setting up libcupsfilters1:amd64 (1.20.2-0ubuntu3.1) ...
Setting up libcupsimage2:amd64 (2.2.7-1ubuntu2.8) ...
Setting up libgs9:amd64 (9.26~dfsg+0-0ubuntu0.18.04.12) ...
Setting up lmodern (2.004.5-3) ...
Setting up texlive-binaries (2017.20170613.44572-8ubuntu0.1) ...
update-alternatives: using /usr/bin/xdvi-xaw to provide /usr/bin/xdvi.bin
(xdvi.bin) in auto mode
update-alternatives: using /usr/bin/bibtex.original to provide /usr/bin/bibtex
(bibtex) in auto mode
Setting up texlive-base (2017.20180305-1) ...
mktexlsr: Updating /var/lib/texmf/ls-R-TEXLIVEDIST...
mktexlsr: Updating /var/lib/texmf/ls-R-TEXMFMAIN...
mktexlsr: Updating /var/lib/texmf/ls-R...
mktexlsr: Done.
tl-paper: setting paper size for dvips to a4: /var/lib/texmf/dvips/config
/config-paper.ps
tl-paper: setting paper size for dvipdfmx to a4: /var/lib/texmf/dvipdfmx
/dvipdfmx-paper.cfg
tl-paper: setting paper size for xdvi to a4: /var/lib/texmf/xdvi/XDvi-paper
tl-paper: setting paper size for pdftex to a4:
/var/lib/texmf/tex/generic/config/pdftexconfig.tex
Setting up texlive-fonts-recommended (2017.20180305-1) ...
Setting up texlive-plain-generic (2017.20180305-2) ...
Setting up texlive-latex-base (2017.20180305-1) ...
Setting up texlive-latex-recommended (2017.20180305-1) ...
Setting up texlive-pictures (2017.20180305-1) ...
Setting up tipa (2:1.3-20) ...
Regenerating '/var/lib/texmf/fmtutil.cnf-DEBIAN'... done.

```

Regenerating '/var/lib/texmf/fmtutil.cnf-TEXLIVEDIST'... done.

update-fmtutil has updated the following file(s):

/var/lib/texmf/fmtutil.cnf-DEBIAN

/var/lib/texmf/fmtutil.cnf-TEXLIVEDIST

If you want to activate the changes in the above file(s),  
you should run fmtutil-sys or fmtutil.

Setting up texlive (2017.20180305-1) ...

Setting up texlive-latex-extra (2017.20180305-2) ...

Setting up texlive-xetex (2017.20180305-1) ...

Setting up ruby2.5 (2.5.1-1ubuntu1.6) ...

Setting up ruby (1:2.5.1) ...

Setting up ruby-test-unit (3.2.5-1) ...

Setting up rake (12.3.1-1ubuntu0.1) ...

Setting up libruby2.5:amd64 (2.5.1-1ubuntu1.6) ...

Processing triggers for mime-support (3.60ubuntu1) ...

Processing triggers for libc-bin (2.27-3ubuntu1) ...

/sbin/ldconfig.real: /usr/local/lib/python3.6/dist-  
packages/ideep4py/lib/libmkldnn.so.0 is not a symbolic link

Processing triggers for man-db (2.8.3-2ubuntu0.1) ...

Processing triggers for fontconfig (2.12.6-0ubuntu2) ...

Processing triggers for tex-common (6.09) ...

Running updmap-sys. This may take some time... done.

Running mktexlsr /var/lib/texmf ... done.

Building format(s) --all.

This may take some time... done.

Requirement already satisfied: py pandoc in /usr/local/lib/python3.6/dist-  
packages (1.5)

Requirement already satisfied: setuptools in /usr/local/lib/python3.6/dist-  
packages (from py pandoc) (46.1.3)

Requirement already satisfied: wheel>=0.25.0 in /usr/local/lib/python3.6/dist-  
packages (from py pandoc) (0.34.2)

Requirement already satisfied: pip>=8.1.0 in /usr/local/lib/python3.6/dist-  
packages (from py pandoc) (19.3.1)

[161]: *# Export the notebook as pdf*  
!sudo apt-get update

Get:1 https://cloud.r-project.org/bin/linux/ubuntu bionic-cran35/ InRelease  
[3,626 B]

Ign:2 https://developer.download.nvidia.com/compute/cuda/repos/ubuntu1804/x86\_64  
InRelease

Get:3 https://cloud.r-project.org/bin/linux/ubuntu bionic-cran35/ Packages [91.7  
kB]

Ign:4 https://developer.download.nvidia.com/compute/machine-  
learning/repos/ubuntu1804/x86\_64 InRelease

Hit:5 https://developer.download.nvidia.com/compute/cuda/repos/ubuntu1804/x86\_64

```

Release
Hit:6 https://developer.download.nvidia.com/compute/machine-
learning/repos/ubuntu1804/x86_64 Release
Get:8 http://ppa.launchpad.net/graphics-drivers/ppa/ubuntu bionic InRelease
[21.3 kB]
Get:10 http://security.ubuntu.com/ubuntu bionic-security InRelease [88.7 kB]
Hit:11 http://archive.ubuntu.com/ubuntu bionic InRelease
Get:12 http://archive.ubuntu.com/ubuntu bionic-updates InRelease [88.7 kB]
Get:13 http://ppa.launchpad.net/marutter/c2d4u3.5/ubuntu bionic InRelease [15.4
kB]
Get:14 http://security.ubuntu.com/ubuntu bionic-security/main amd64 Packages
[908 kB]
Get:15 http://ppa.launchpad.net/graphics-drivers/ppa/ubuntu bionic/main amd64
Packages [37.4 kB]
Get:16 http://archive.ubuntu.com/ubuntu bionic-backports InRelease [74.6 kB]
Get:17 http://security.ubuntu.com/ubuntu bionic-security/restricted amd64
Packages [52.4 kB]
Get:18 http://security.ubuntu.com/ubuntu bionic-security/universe amd64 Packages
[844 kB]
Get:19 http://security.ubuntu.com/ubuntu bionic-security/multiverse amd64
Packages [8,505 B]
Get:20 http://archive.ubuntu.com/ubuntu bionic-updates/restricted amd64 Packages
[66.8 kB]
Get:21 http://archive.ubuntu.com/ubuntu bionic-updates/universe amd64 Packages
[1,376 kB]
Get:22 http://ppa.launchpad.net/marutter/c2d4u3.5/ubuntu bionic/main Sources
[1,813 kB]
Get:23 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 Packages
[1,205 kB]
Get:24 http://archive.ubuntu.com/ubuntu bionic-updates/multiverse amd64 Packages
[19.8 kB]
Get:25 http://archive.ubuntu.com/ubuntu bionic-backports/main amd64 Packages
[8,286 B]
Get:26 http://archive.ubuntu.com/ubuntu bionic-backports/universe amd64 Packages
[8,158 B]
Get:27 http://ppa.launchpad.net/marutter/c2d4u3.5/ubuntu bionic/main amd64
Packages [875 kB]
Fetched 7,606 kB in 7s (1,114 kB/s)
Reading package lists... Done
[NbConvertApp] Converting notebook ./gdrive/My Drive/Colab
Notebooks/MAEG5735-2020-Assignment4/FinalAssignment_S1155135359.ipynb to PDF
[NbConvertApp] Support files will be in FinalAssignment_S1155135359_files/
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Writing 164317 bytes to ./notebook.tex

```

```

[NbConvertApp] Building PDF
Traceback (most recent call last):
  File "/usr/local/bin/jupyter-nbconvert", line 8, in <module>
    sys.exit(main())
  File "/usr/local/lib/python2.7/dist-packages/jupyter_core/application.py",
line 267, in launch_instance
    return super(JupyterApp, cls).launch_instance(argv=argv, **kwargs)
  File "/usr/local/lib/python2.7/dist-packages/traitlets/config/application.py",
line 658, in launch_instance
    app.start()
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/nbconvertapp.py", line
338, in start
    self.convert_notebooks()
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/nbconvertapp.py", line
508, in convert_notebooks
    self.convert_single_notebook(notebook_filename)
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/nbconvertapp.py", line
479, in convert_single_notebook
    output, resources = self.export_single_notebook(notebook_filename,
resources, input_buffer=input_buffer)
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/nbconvertapp.py", line
408, in export_single_notebook
    output, resources = self.exporter.from_filename(notebook_filename,
resources=resources)
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/exporters/exporter.py",
line 179, in from_filename
    return self.from_file(f, resources=resources, **kw)
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/exporters/exporter.py",
line 197, in from_file
    return self.from_notebook_node(nbformat.read(file_stream, as_version=4),
resources=resources, **kw)
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/exporters/pdf.py", line
178, in from_notebook_node
    rc = self.run_latex(tex_file)
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/exporters/pdf.py", line
149, in run_latex
    self.latex_count, log_error)
  File "/usr/local/lib/python2.7/dist-packages/nbconvert/exporters/pdf.py", line
111, in run_command
    "at {link}.".format(formatter=command_list[0], link=link))
OSError: xelatex not found on PATH, if you have not installed xelatex you may
need to do so. Find further instructions at
https://nbconvert.readthedocs.io/en/latest/install.html#installing-tex.

```

[163]: !jupyter nbconvert --to PDF "./gdrive/My Drive/Colab Notebooks/  
↪MAEG5735-2020-Assignment4/FinalAssignment\_S1155135359.ipynb"

[NbConvertApp] Converting notebook ./gdrive/My Drive/Colab

```
Notebooks/MAEG5735-2020-Assignment4/FinalAssignment_S1155135359.ipynb to PDF
[NbConvertApp] Support files will be in FinalAssignment_S1155135359_files/
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Making directory ./FinalAssignment_S1155135359_files
[NbConvertApp] Writing 164317 bytes to ./notebook.tex
[NbConvertApp] Building PDF
[NbConvertApp] Running xelatex 3 times: [u'xelatex', u'./notebook.tex',
'-quiet']
[NbConvertApp] Running bibtex 1 time: [u'bibtex', u'./notebook']
[NbConvertApp] WARNING | bibtex had problems, most likely because there were no
citations
[NbConvertApp] PDF successfully created
[NbConvertApp] Writing 203548 bytes to ./gdrive/My Drive/Colab
Notebooks/MAEG5735-2020-Assignment4/FinalAssignment_S1155135359.pdf
```