# pytorch\_cl\_vae\_train

September 15, 2018

# 1 Basic Classifying VAE for MNIST Database

### 1.1 1 - Specify parameters for to the VAE and training

#### 1.2 2 - Fetch MNIST

```
lb.fit(mnist.target)
        params['classes_dim'] = [num_classes]
        params['original_dim'] = input_dim
        print('MNIST db has been successfully loaded, stored in the: "{}"'.format(params['data
        # split data to train and test subsets
        X_train, X_test, y_train, y_test = train_test_split(mnist.data, mnist.target, test_size
        print("| Train subset shape:{} | Test subset shape:{} | ".format(X_train.shape, X_test.;
MNIST db has been successfully loaded, stored in the: "../data/mldata"
| Train subset shape: (63000, 784) | Test subset shape: (7000, 784) |
1.3 3 - Create Model
In [4]: # Initialize ClVaeModel
        model = ClVaeModel(**params)
        print("Model successfully initialized with params: ")
        pprint.PrettyPrinter(indent=4).pprint(params)
        train_losses = []
        train_accuracies = []
Model successfully initialized with params:
    'batch_size': 100,
    'classes_dim': [10],
    'classifier_hidden_size': 512,
    'classifier_learning_rate': 0.0001,
    'data_dir': '../data',
    'decoder_hidden_size': 512,
    'encoder_hidden_size': 512,
    'latent_dim': 2,
    'log_dir': '../data/logs',
    'model_dir': '../data/models',
    'num_epochs': 5,
    'original_dim': 784,
    'vae_learning_rate': 0.0001}
1.4 4 - Train
In [5]: save_each_steps = 500
        # Train loop
        train_step_i = 0
        for epoch in range(params['num_epochs']):
            print('\nepoch {} out of {}'.format(epoch + 1, params['num_epochs']))
            for i in range(X_train.shape[0] // params['batch_size']):
                # Sample batch
```

```
x_batch = torch.from_numpy(X_train[idx]).float()
                                                    y_batch = lb.transform(y_train[idx])
                                                    y_batch = [torch.from_numpy(y_batch).float()]
                                                    step losses, step accuracies = model.train step(x batch, y batch)
                                                           step_losses = [loss.sum().detach().numpy() for loss in step_losses]
                                                     # step_losses = Losses(*step_losses)
                                                     # step accuracies = Accuracies(*step accuracies)
                                                    train_losses.append(step_losses)
                                                    train_accuracies.append(step_accuracies)
                                                    train_step_i += 1
                                                    print("\r|train step: {} | rec loss: {:.4f} | z_dkl loss: {:.4f} | class loss:
                                                                         " | w_dkl loss: {:.4f} | class_accuracy: {:.4f} |".format(
                                                                 train_step_i, *step_losses, *step_accuracies
                                                                  ), end='')
                                                    if train_step_i % 100 == 0:
                                                                 print()
                                                    if train_step_i % save_each_steps == 0:
                                                                 dt = str(datetime.datetime.now().strftime("%m_%d_%Y_%I_%M_%p"))
                                                                 fname = params['model_dir'] + '/cl_vae_mnist_{}.pt'.format(dt)
                                                                 model.save_ckpt(fname)
                          print('*****Finished with the final loss: ', step_losses)
epoch 1 out of 5
|train step: 100 | rec loss: 0.4931 | z_dkl loss: 0.0012 | class loss: 2.3088 | w_dkl loss: 0.0012 | class loss: 2.3088 | class loss: 2
|train step: 200 | rec loss: 0.3779 | z dkl loss: 0.0010 | class loss: 2.2954 | w dkl loss: 0.0010 |
|train step: 300 | rec loss: 0.3180 | z_dkl loss: 0.0006 | class loss: 2.2958 | w_dkl loss: 0.0006
|train step: 400 | rec loss: 0.2896 | z_dkl loss: 0.0004 | class loss: 2.2972 | w_dkl loss: 0.0004 | class loss: 2.2972 | class loss: 2
|train step: 500 | rec loss: 0.2736 | z_dkl loss: 0.0003 | class loss: 2.2923 | w_dkl loss: 0.0003 |
|train step: 600 | rec loss: 0.2611 | z_dkl loss: 0.0002 | class loss: 2.3019 | w_dkl loss: 0.0002 |
|train step: 630 | rec loss: 0.2530 | z_dkl loss: 0.0002 | class loss: 2.3122 | w_dkl loss: 0.0002 |
epoch 2 out of 5
|train step: 700 | rec loss: 0.2491 | z_dkl loss: 0.0002 | class loss: 2.2944 | w_dkl loss: 0.0002 |
|train step: 800 | rec loss: 0.2502 | z_dkl loss: 0.0002 | class loss: 2.3092 | w_dkl loss: 0.0002 | class loss: 2.3092 | class loss: 2
|train step: 900 | rec loss: 0.2302 | z_dkl loss: 0.0002 | class loss: 2.3001 | w_dkl loss: 0.0002
|train step: 1000 | rec loss: 0.2307 | z_dkl loss: 0.0001 | class loss: 2.3042 | w_dkl loss: 0
|train step: 1100 | rec loss: 0.2311 | z_dkl loss: 0.0001 | class loss: 2.2950 | w_dkl loss: 0
|train step: 1200 | rec loss: 0.2292 | z_dkl loss: 0.0001 | class loss: 2.3217 | w_dkl loss: 0
|train step: 1260 | rec loss: 0.2396 | z dkl loss: 0.0001 | class loss: 2.3059 | w dkl loss: 0
epoch 3 out of 5
|train step: 1300 | rec loss: 0.2206 | z_dkl loss: 0.0001 | class loss: 2.3141 | w_dkl loss: 0
|train step: 1400 | rec loss: 0.2380 | z dkl loss: 0.0001 | class loss: 2.3149 | w dkl loss: 0
|train step: 1500 | rec loss: 0.2293 | z_dkl loss: 0.0001 | class loss: 2.3063 | w_dkl loss: 0
```

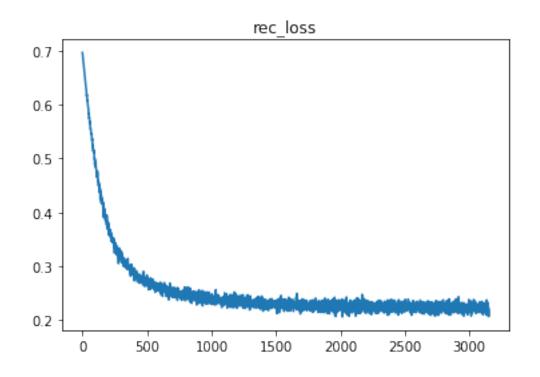
idx = random.choice(np.arange(0, X\_train.shape[0]), params['batch\_size'])

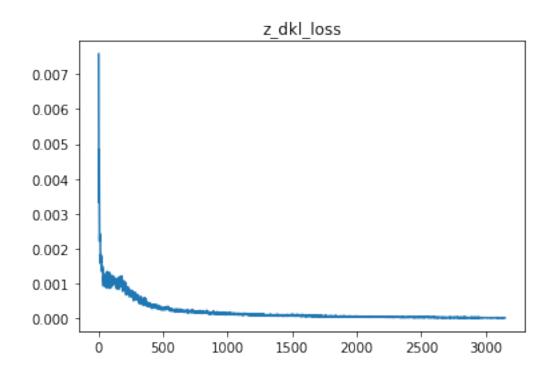
```
|train step: 1600 | rec loss: 0.2358 | z dkl loss: 0.0001 | class loss: 2.3002 | w dkl loss: 0
|train step: 1700 | rec loss: 0.2248 | z_dkl loss: 0.0001 | class loss: 2.3042 | w_dkl loss: 0
|train step: 1800 | rec loss: 0.2360 | z dkl loss: 0.0001 | class loss: 2.2923 | w dkl loss: 0
|train step: 1890 | rec loss: 0.2327 | z_dkl loss: 0.0001 | class loss: 2.3046 | w_dkl loss: 0
epoch 4 out of 5
|train step: 1900 | rec loss: 0.2120 | z_dkl loss: 0.0001 | class loss: 2.2911 | w_dkl loss: 0
|train step: 2000 | rec loss: 0.2325 | z dkl loss: 0.0001 | class loss: 2.3158 | w dkl loss: 0
|train step: 2100 | rec loss: 0.2322 | z_dkl loss: 0.0001 | class loss: 2.3010 | w_dkl loss: 0
|train step: 2200 | rec loss: 0.2280 | z_dkl loss: 0.0001 | class loss: 2.3139 | w_dkl loss: 0
|train step: 2300 | rec loss: 0.2239 | z_dkl loss: 0.0000 | class loss: 2.2909 | w_dkl loss: 0
|train step: 2400 | rec loss: 0.2214 | z_dkl loss: 0.0000 | class loss: 2.3045 | w_dkl loss: 0
|train step: 2500 | rec loss: 0.2148 | z dkl loss: 0.0000 | class loss: 2.2873 | w dkl loss: 0
|train step: 2520 | rec loss: 0.2207 | z_dkl loss: 0.0000 | class loss: 2.3018 | w_dkl loss: 0
epoch 5 out of 5
|train step: 2600 | rec loss: 0.2199 | z_dkl loss: 0.0000 | class loss: 2.2972 | w_dkl loss: 0
|train step: 2700 | rec loss: 0.2264 | z dkl loss: 0.0000 | class loss: 2.2913 | w dkl loss: 0
|train step: 2800 | rec loss: 0.2296 | z_dkl loss: 0.0000 | class loss: 2.2997 | w_dkl loss: 0
|train step: 2900 | rec loss: 0.2179 | z dkl loss: 0.0000 | class loss: 2.3035 | w dkl loss: 0
|train step: 3000 | rec loss: 0.2240 | z_dkl loss: 0.0000 | class loss: 2.3218 | w_dkl loss: 0
|train step: 3100 | rec loss: 0.2287 | z dkl loss: 0.0000 | class loss: 2.3018 | w dkl loss: 0
|train step: 3150 | rec loss: 0.2070 | z_dkl loss: 0.0000 | class loss: 2.2842 | w_dkl loss: 0
```

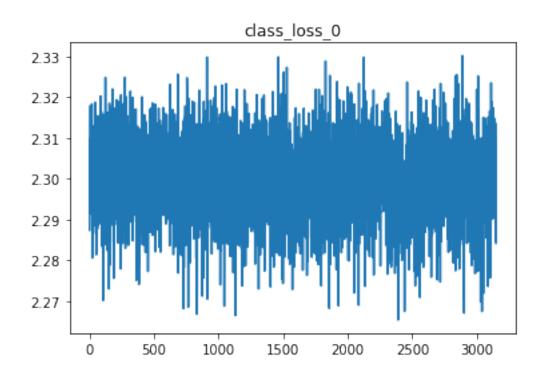
#### 1.5 5 - Show losses graph

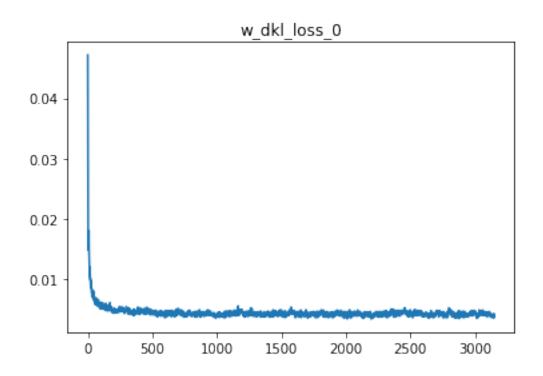
```
In [6]: %matplotlib inline
    losses = defaultdict(list)
    losses_names = train_losses[0]._fields
    print(losses_names)
    step_loss = train_losses[0]
    print(*step_loss)
    for i, loss_name in enumerate(losses_names):
        losses[loss_name] = [1[i] for 1 in train_losses]
        plt.figure()
        plt.title(loss_name)
        plt.plot(losses[loss_name])
        plt.legend()
    plt.show()

('rec_loss', 'z_dkl_loss', 'class_loss_0', 'w_dkl_loss_0')
tensor(0.6971, grad_fn=<BinaryCrossEntropyBackward>) tensor(0.0076, grad_fn=<DivBackward0>) tensor(0.0076, grad_fn=<DivBackward0>) tensor(0.0076, grad_fn=<DivBackward0>) tensor(0.0076, grad_fn=<DivBackward0>) tensor(0.0076, grad_fn=<DivBackward0>)
```









### 1.6 6 - Test

Losses(rec\_loss=tensor(0.2668), z\_dkl\_loss=tensor(0.0000), class\_loss\_0=tensor(2.3001), w\_dkl\_i

## 1.7 7 - Examples

In [ ]: #TODO