

# diffusion\_equation

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## 1 Using fluidlearn to solve diffusion equation

Equation to solve:  $u_t - \Delta u - f = 0$  over domain  $\Omega$  from time  $T_{\text{initial}}$  to  $T_{\text{final}}$ .

For demonstration purposes we take  $f = \sin(x_1 + x_2) + t\sin(x_1 + x_2)$  and  $\Omega = [-2, 2] \times [0, 1]$  and the time interval to be  $[0, 1]$ , so we can compare the results with the actual solution  $u = t\sin(x_1 + x_2)$ .

```
[1]: #Import fluidlearn package and classes
import fluidlearn
from fluidlearn import dataprocess
```

### 1.0.1 Defining the domain and time interval for which the PDE needs to be solved.

This matters only for generating collocation points and if the user is feeding their own collocation points, they can skip this step.

```
[2]: #domain range
X_1_domain = [-2, 2]
X_2_domain = [0, 1]
#time range
T_initial = 0
T_final = 1
T_domain = [T_initial, T_final]

#domain of the problem
domain_bounds = [X_1_domain, X_2_domain, T_domain]
```

### 1.0.2 Loading data from a csv file

- We use the manufactured data with  $u = t\sin(x_1 + x_2)$  saved in a csv file.
- Data is saved in the format:  $(x_1, x_2, t, u(x_1, x_2, t))$  as four columns.
- You could load either preprocess your data to be in this format or load your data from a csv file with similar format.

```
[3]: path_to_data = "data_manufactured/t_sin_x_plus_y.csv"
X_data, Y_data = dataprocess.imp_from_csv(path_to_csv_file=path_to_data,
                                           x_y_combined=True, y_dim=1)
```

### 1.0.3 Defining the rhs function $f = \sin(x_1 + x_2) + t\sin(x_1 + x_2)$ of the PDE.

We use tensorflow.sin function instead of python functions, we could used numpy.sin as well.

```
[4]: def rhs_function (args, time_dep=True):
      import tensorflow as tf
      if time_dep:
          space_inputs = args[:-1]
          time_inputs = args[-1]
      else:
          space_inputs = args

      return tf.sin(space_inputs[0]+space_inputs[1]) + 2*time_inputs*tf.
      ↪sin(space_inputs[0]+space_inputs[1])
```

### 1.0.4 Defining the model architecture

```
[5]: model_type = 'forward'
      space_dim = 2 #dimension of Omega
      time_depedent_problem = True
      n_hid_lay=3 #numberof hidden layers in the neural network
      n_hid_nrn=20 #number of neurons in each hidden layer
      act_func='tanh' #activation function used for hidden layers: could be elu, ↪
      ↪relu, sigmoid
      loss_list='mse' #type of error function used for cost functin, we use mean ↪
      ↪squared error.
      optimizer='adam' #type of optimizer for cost function minimization
      dom_bounds=domain_bounds #domain bounds where collocation points has to be ↪
      ↪generated

      distribution = 'uniform' #type of distribution used for generating the pde ↪
      ↪collocation points.
      number_of_collocation_points = 5000

      batch_size = 32 #batch size for stochastic batch gradient type optimization
      num_epochs = 10 #number of epochs used for trainng
```

### 1.0.5 Defining the fluidlearn solver

```
[41]: diffusion_model = fluidlearn.Solver()
```

```
[42]: diffusion_model(model_type=model_type,
                      space_dim=space_dim,
                      time_dep=time_depedent_problem,
                      output_dim=1,
                      n_hid_lay=n_hid_lay,
                      n_hid_nrn=n_hid_lay,
```

```

        act_func=act_func,
        rhs_func=rhs_function,
        loss_list=loss_list,
        optimizer=optimizer,
        dom_bounds=dom_bounds,
        load_model=False,
        model_path=None,)

```

### 1.0.6 Fitting the model

```

[44]: diffusion_model.fit(
        x=X_data,
        y=Y_data,
        colloc_points=number_of_collocation_points,
        dist=distribution,
        batch_size=batch_size,
        epochs=num_epochs,
    )

```

```

Epoch 1/10
219/219 [=====] - 0s 2ms/step - loss: 0.2146 -
output_1_loss: 0.0297 - output_2_loss: 0.1849
Epoch 2/10
219/219 [=====] - 0s 2ms/step - loss: 0.1877 -
output_1_loss: 0.0355 - output_2_loss: 0.1521
Epoch 3/10
219/219 [=====] - 0s 2ms/step - loss: 0.1598 -
output_1_loss: 0.0388 - output_2_loss: 0.1210
Epoch 4/10
219/219 [=====] - 0s 2ms/step - loss: 0.1298 -
output_1_loss: 0.0367 - output_2_loss: 0.0931
Epoch 5/10
219/219 [=====] - 0s 1ms/step - loss: 0.0966 -
output_1_loss: 0.0304 - output_2_loss: 0.0662
Epoch 6/10
219/219 [=====] - 0s 2ms/step - loss: 0.0709 -
output_1_loss: 0.0228 - output_2_loss: 0.0480
Epoch 7/10
219/219 [=====] - 0s 2ms/step - loss: 0.0572 -
output_1_loss: 0.0194 - output_2_loss: 0.0378
Epoch 8/10
219/219 [=====] - 0s 2ms/step - loss: 0.0490 -
output_1_loss: 0.0179 - output_2_loss: 0.0311
Epoch 9/10
219/219 [=====] - 0s 2ms/step - loss: 0.0436 -
output_1_loss: 0.0175 - output_2_loss: 0.0262
Epoch 10/10

```

```
219/219 [=====] - 0s 2ms/step - loss: 0.0402 -  
output_1_loss: 0.0166 - output_2_loss: 0.0235
```

### 1.0.7 Resuming Training the model again for 50 more epochs

```
[45]: diffusion_model.fit(  
    x=X_data,  
    y=Y_data,  
    colloc_points=number_of_collocation_points,  
    dist=distribution,  
    batch_size=batch_size,  
    epochs=50,  
)
```

Epoch 1/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0367 -  
output_1_loss: 0.0160 - output_2_loss: 0.0207
```

Epoch 2/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0343 -  
output_1_loss: 0.0150 - output_2_loss: 0.0193
```

Epoch 3/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0320 -  
output_1_loss: 0.0145 - output_2_loss: 0.0176
```

Epoch 4/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0304 -  
output_1_loss: 0.0138 - output_2_loss: 0.0166
```

Epoch 5/50

```
219/219 [=====] - 0s 1ms/step - loss: 0.0286 -  
output_1_loss: 0.0129 - output_2_loss: 0.0157
```

Epoch 6/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0270 -  
output_1_loss: 0.0122 - output_2_loss: 0.0148
```

Epoch 7/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0254 -  
output_1_loss: 0.0113 - output_2_loss: 0.0141
```

Epoch 8/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0242 -  
output_1_loss: 0.0106 - output_2_loss: 0.0135
```

Epoch 9/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0226 -  
output_1_loss: 0.0099 - output_2_loss: 0.0127
```

Epoch 10/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0213 -  
output_1_loss: 0.0090 - output_2_loss: 0.0122
```

Epoch 11/50

```
219/219 [=====] - 0s 2ms/step - loss: 0.0198 -  
output_1_loss: 0.0083 - output_2_loss: 0.0115
```

Epoch 12/50

219/219 [=====] - 0s 2ms/step - loss: 0.0183 -  
output\_1\_loss: 0.0075 - output\_2\_loss: 0.0108  
Epoch 13/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0168 -  
output\_1\_loss: 0.0066 - output\_2\_loss: 0.0101  
Epoch 14/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0148 -  
output\_1\_loss: 0.0056 - output\_2\_loss: 0.0092  
Epoch 15/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0129 -  
output\_1\_loss: 0.0048 - output\_2\_loss: 0.0081  
Epoch 16/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0109 -  
output\_1\_loss: 0.0038 - output\_2\_loss: 0.0070  
Epoch 17/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0089 -  
output\_1\_loss: 0.0031 - output\_2\_loss: 0.0058  
Epoch 18/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0070 -  
output\_1\_loss: 0.0025 - output\_2\_loss: 0.0046  
Epoch 19/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0056 -  
output\_1\_loss: 0.0020 - output\_2\_loss: 0.0036  
Epoch 20/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0047 -  
output\_1\_loss: 0.0018 - output\_2\_loss: 0.0029  
Epoch 21/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0039 -  
output\_1\_loss: 0.0016 - output\_2\_loss: 0.0023  
Epoch 22/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0035 -  
output\_1\_loss: 0.0015 - output\_2\_loss: 0.0020  
Epoch 23/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0031 -  
output\_1\_loss: 0.0014 - output\_2\_loss: 0.0017  
Epoch 24/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0028 -  
output\_1\_loss: 0.0013 - output\_2\_loss: 0.0015  
Epoch 25/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0025 -  
output\_1\_loss: 0.0012 - output\_2\_loss: 0.0013  
Epoch 26/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0023 -  
output\_1\_loss: 0.0011 - output\_2\_loss: 0.0012  
Epoch 27/50  
219/219 [=====] - 0s 2ms/step - loss: 0.0021 -  
output\_1\_loss: 0.0011 - output\_2\_loss: 0.0011  
Epoch 28/50

219/219 [=====] - 0s 2ms/step - loss: 0.0019 -  
 output\_1\_loss: 9.6183e-04 - output\_2\_loss: 9.5103e-04  
 Epoch 29/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0017 -  
 output\_1\_loss: 9.0621e-04 - output\_2\_loss: 8.3377e-04  
 Epoch 30/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0016 -  
 output\_1\_loss: 8.5757e-04 - output\_2\_loss: 7.6970e-04  
 Epoch 31/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0015 -  
 output\_1\_loss: 8.0180e-04 - output\_2\_loss: 6.7573e-04  
 Epoch 32/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0014 -  
 output\_1\_loss: 7.5866e-04 - output\_2\_loss: 6.2616e-04  
 Epoch 33/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0013 -  
 output\_1\_loss: 7.2239e-04 - output\_2\_loss: 5.5569e-04  
 Epoch 34/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0012 -  
 output\_1\_loss: 6.8165e-04 - output\_2\_loss: 4.9617e-04  
 Epoch 35/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0011 -  
 output\_1\_loss: 6.4506e-04 - output\_2\_loss: 4.5623e-04  
 Epoch 36/50  
 219/219 [=====] - 0s 2ms/step - loss: 0.0011 -  
 output\_1\_loss: 6.2371e-04 - output\_2\_loss: 4.3896e-04  
 Epoch 37/50  
 219/219 [=====] - 0s 2ms/step - loss: 9.8307e-04 -  
 output\_1\_loss: 5.9644e-04 - output\_2\_loss: 3.8664e-04  
 Epoch 38/50  
 219/219 [=====] - 0s 2ms/step - loss: 9.4429e-04 -  
 output\_1\_loss: 5.7172e-04 - output\_2\_loss: 3.7257e-04  
 Epoch 39/50  
 219/219 [=====] - 0s 2ms/step - loss: 8.8748e-04 -  
 output\_1\_loss: 5.3745e-04 - output\_2\_loss: 3.5004e-04  
 Epoch 40/50  
 219/219 [=====] - 0s 2ms/step - loss: 8.6691e-04 -  
 output\_1\_loss: 5.1713e-04 - output\_2\_loss: 3.4978e-04  
 Epoch 41/50  
 219/219 [=====] - 0s 2ms/step - loss: 8.1541e-04 -  
 output\_1\_loss: 4.9690e-04 - output\_2\_loss: 3.1851e-04  
 Epoch 42/50  
 219/219 [=====] - 0s 2ms/step - loss: 7.9440e-04 -  
 output\_1\_loss: 4.8338e-04 - output\_2\_loss: 3.1103e-04  
 Epoch 43/50  
 219/219 [=====] - 0s 2ms/step - loss: 8.1476e-04 -  
 output\_1\_loss: 4.7769e-04 - output\_2\_loss: 3.3707e-04  
 Epoch 44/50

```

219/219 [=====] - 0s 2ms/step - loss: 7.8507e-04 -
output_1_loss: 4.7123e-04 - output_2_loss: 3.1384e-04
Epoch 45/50
219/219 [=====] - 0s 2ms/step - loss: 7.6107e-04 -
output_1_loss: 4.5173e-04 - output_2_loss: 3.0934e-04
Epoch 46/50
219/219 [=====] - 0s 2ms/step - loss: 7.3157e-04 -
output_1_loss: 4.3707e-04 - output_2_loss: 2.9450e-04
Epoch 47/50
219/219 [=====] - 0s 2ms/step - loss: 7.1687e-04 -
output_1_loss: 4.2378e-04 - output_2_loss: 2.9309e-04
Epoch 48/50
219/219 [=====] - 0s 2ms/step - loss: 7.0904e-04 -
output_1_loss: 4.2875e-04 - output_2_loss: 2.8029e-04
Epoch 49/50
219/219 [=====] - 0s 2ms/step - loss: 6.8987e-04 -
output_1_loss: 4.0897e-04 - output_2_loss: 2.8089e-04
Epoch 50/50
219/219 [=====] - 0s 2ms/step - loss: 6.7420e-04 -
output_1_loss: 4.0158e-04 - output_2_loss: 2.7262e-04

```

### 1.0.8 Saving the model to a specified location.

```

[48]: path_to_save_model = "saved_model/model_name"
diffusion_model.save_model(path_to_save_model)

```

```

WARNING:tensorflow:From C:\Users\manuj\anaconda3\envs\tf2\lib\site-
packages\tensorflow\python\training\ttracking\ttracking.py:111:
Model.state_updates (from tensorflow.python.keras.engine.training) is deprecated
and will be removed in a future version.
Instructions for updating:
This property should not be used in TensorFlow 2.0, as updates are applied
automatically.
WARNING:tensorflow:From C:\Users\manuj\anaconda3\envs\tf2\lib\site-
packages\tensorflow\python\training\ttracking\ttracking.py:111: Layer.updates
(from tensorflow.python.keras.engine.base_layer) is deprecated and will be
removed in a future version.
Instructions for updating:
This property should not be used in TensorFlow 2.0, as updates are applied
automatically.
INFO:tensorflow:Assets written to: saved_model/model_name/assets

```

### 1.0.9 Loading the saved model

```

[6]: path_to_load_model = "saved_model/model_name"

```

```

[7]: loaded_diffusion_model = fluidlearn.Solver()

```

```
[8]: loaded_diffusion_model(space_dim=2,  
    time_dep=True,  
    load_model=True,  
    model_path=path_to_load_model)
```

#### 1.0.10 Predicting using loaded model

```
[9]: y_predicted = loaded_diffusion_model.predict(X_data)
```

```
[10]: y_predicted
```

```
[10]: array([[ -0.10157388],  
    [-0.4190994 ],  
    [-0.7965628 ],  
    ...,  
    [-0.10375804],  
    [ 0.05802408],  
    [-0.00470909]], dtype=float32)
```

```
[ ]:
```