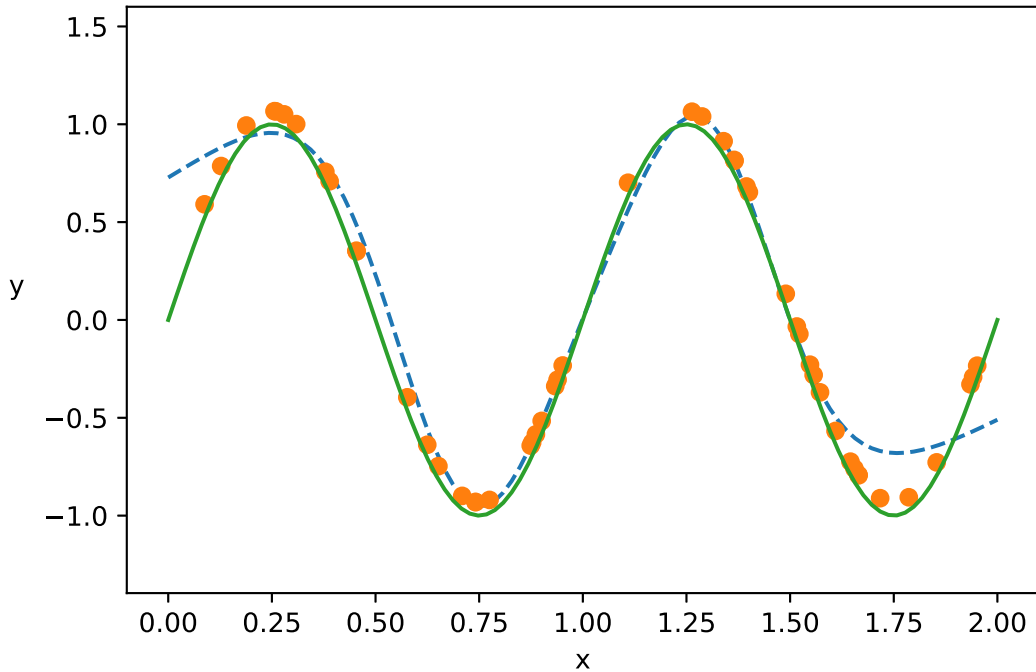


Sep 07, 22 21:18	main.py	Page 1/2
<pre>#!/bin/env python3.8  """ Example assignment. Author: Chris Curro """ import os  # from turtle import shape import matplotlib.pyplot as plt import numpy as np import tensorflow as tf import math  from tqdm import trange  script_path = os.path.dirname(os.path.realpath(__file__)) rng = np.random.default_rng(seed=42)  class Data:     def __init__(self, ns, sig, range):         self.rng = np.random.default_rng(seed=42)         self.index = np.arange(ns)         self.num_samples = ns         self.sigma = sig         self.range = range         self.x = np.array(self.rng.uniform(0, 2, self.num_samples))         clean_y = tf.math.sin(2 * math.pi * self.x)         self.y = np.array(clean_y + self.rng.normal(0, self.sigma))      def get_batch(self, batch_size):         choices = np.array(rng.choice(self.index, size=batch_size))         return self.x[choices], self.y[choices].flatten()  class Model(tf.Module):     def __init__(self, M):         self.M = M         self.w = tf.Variable(rng.normal(0, 1, M))         self.mu = tf.Variable(rng.normal(1, 0.5, M))         self.sigma = tf.Variable(rng.normal(0, 1, M))         self.b = tf.Variable(np.array([[0.0]]))      def __call__(self, x):         num = tf.shape(x)[0]         x = tf.squeeze(x)         x = tf.transpose(tf.broadcast_to(x, [self.M, num]))         mu = tf.broadcast_to(self.mu, [num, self.M])         sigma = tf.broadcast_to(self.sigma, [num, self.M])         theta = tf.math.exp(-(x - mu) ** 2) / (sigma ** 2)          y = tf.broadcast_to(self.w, [1, self.M]) @ tf.transpose(theta)          return tf.cast(y, dtype="float32") + tf.cast(self.b, dtype="float32")  data = Data(50, 0.1, (0, 2)) model = Model(5)  optimizer = tf.optimizers.SGD(learning_rate=0.1)</pre>		

Sep 07, 22 21:18	main.py	Page 2/2
<pre>bar = trange(500) for i in bar:     with tf.GradientTape() as tape:         x, y = data.get_batch(16)         y_hat = model(x)         loss = 0.5 * tf.reduce_mean((y_hat - y) ** 2)          grads = tape.gradient(loss, model.trainable_variables)         optimizer.apply_gradients(zip(grads, model.trainable_variables))      bar.set_description(f"Loss @ {i} =&gt; {loss.numpy():0.6f}")     bar.refresh()  fig, ax = plt.subplots(1, 2, figsize=(11, 4), dpi=200)  ax[0].set_title("Linear Combination of Gaussians") ax[0].set_xlabel("x") ax[0].set_ylim(np.amin(data.y) * 1.5, np.amax(data.y) * 1.5) h = ax[0].set_ylabel("y", labelpad=10) h.set_rotation(0)  xs = np.linspace(0, 2, 100) xs = xs[:, np.newaxis] print(tf.shape(np.squeeze(xs))) print(tf.shape(model(np.squeeze(xs)))) ax[0].plot(xs, np.squeeze(model(xs)), "--", label="model") ax[0].plot(np.squeeze(data.x), data.y, "o", label="training data") ax[0].plot(xs, np.squeeze((tf.math.sin(2 * math.pi * xs))), label="training data") ax[0].plot()  ax[1].set_title("Linear Combination of Gaussians") ax[1].set_xlabel("x") ax[1].set_ylim(np.amin(data.y) * 1.5, np.amax(data.y) * 1.5)  for mu_i in range(tf.shape(model.mu)[0]):     theta_j = tf.math.exp(-(xs - model.mu[mu_i]) ** 2) / (model.sigma[mu_i] **     2)     ax[1].plot(xs, theta_j) ax[0].plot() ax[1].plot()  plt.tight_layout() plt.savefig(f"{script_path}/fit.pdf")</pre>		

# Linear Combination of Gaussians



# Linear Combination of Gaussians

