

Load Dataset

Load **diabetes.csv** to two separate numpy arrays **X1**, **y**

Load features into **X1** [Pregnancies, Glucose, BloodPressure, SkinThickness, Insulin, BMI, DiabetesPedigreeFunction, Age]

Load labels into **y** [Outcome 0/1]

Skip the first header row

X1 = (768, 8) # 8 columns with 768 rows

y = (768, 1) # 1 columns with 768 rows

Normalize columns of X1 (Values should be between 0-1):

$$X1 = \frac{X1 - X1.min(axis=0)}{X1.ptp(axis=0)}$$

Augment X at 1st column with a column of 1s:

Add a new column of 1s at first column

X = (768, 9) # 9 columns with 768 rows

Shuffling

Randomize X and corresponding y

Train Test Split

X_train = first 80% of X

y_train = first 80% of y

X_test = rest X

y_test = rest y

X_train = (M, 9) # 9 columns with M rows

y_train = (M, 1) # 1 columns with M rows

X_test = (M', 9) # 9 columns with M' rows

y_test = (M', 1) # 1 columns with M' rows

Initialize θ :

```
lr = 0.01
loss_list = []
total_loss = 0.0
theta_old = new numpy array with 9 random values of shape (9, 1)
theta_new = None
```

Training Loop:

For e in iterations(100):

Calculate Predictions:

```
# matrix multiply X_train and theta_old
Z = matmul( X_train, theta_old ) # Z should be of dimension (M, 1)
h = sigmoid( Z ) # h should be of dimension (M, 1)
```

Calculate Loss values:

```
J_vect = -y_train*log(h) - (1-y_train)*log(1-h) # should be of dimension (M, 1)
J = J_vect.mean() # should be a single numeric value
total_loss += J
loss_list.append(total_loss)
```

Update θ with gradient:

```
grad = numpy.matmul( X_trainT, (h-y_train) )
# grad should be of shape (9, 1) i.e. same as theta_old
theta_new = theta_old - lr * grad
# theta_new should be of shape (9, 1) i.e. same as theta_old

theta_old = theta_new
```

Repeat next iteration**Evaluation on Test Set:**

```
# calculate predictions for X_test
Z = matmul( X_test, theta_old ) # Z should be of dimension (M', 1)
h = sigmoid( Z ) # h should be of dimension (M', 1)
h[h >= 0.5] = 1 # predict 1 where sigmoid value is greater than or equal to 0.5
h[h < 0.5] = 0 # predict 0 where sigmoid value is less than 0.5
```

```
y_test_predicted = h
```

Calculate the accuracy by comparing y_test and y_test_predicted
Print the accuracy (Test)

```
# similarly calculate predictions for X_train too and report
```

Notes:

- Load csv into numpy array
- shuffle two numpy arrays together
- train test split
- Concatenate numpy arrays
- Sigmoid function
- Column wise normalize