Lab 2: Linear Classification and Logistic Regression

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def predict lc(x, theta):
input x: np.ndarray of shape (m, 3)
input theta: np.ndarray of shape (3, 1)
output y: np.ndarray of shape (m, 1)
h = np.dot(x, theta) # dim (m,1)
y = np.sign(h) # dim (m,1)
return y
def mse(y_true, y_pred):
input y_true: np.ndarray of shape (m, 1)
input y_pred: np.ndarray of shape (m, 1)
# Insert your code here ~ 1-3 lines
m = y true.shape[0]
res2 = (y_true - y_pred)**2 # dim (m,1)
J = np.sum(res2) / (2*m)
return J
def gradient(y true, y pred, x):
input y true: np.ndarray of shape (m,)
input y_pred: np.ndarray of shape (m,)
input x: np.ndarray of shape (m, 3)
output dJ: np.array of shape (3, 1)
\mathbf{I} = \mathbf{I} - \mathbf{I}
# Reshape arrays
y true = y true.reshape(-1, 1) # now shape (m, 1)
y pred = y pred.reshape(-1, 1) # now shape (m, 1)
# Your code here ~ 1-6 lines
m = y true.shape[0]
res = y_pred - y_true # dim (m,1)
dJ = np.dot(x.T, res) / m # dim (3,1)
return dJ
```

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def sigmoid(z):
input z: np.ndaray of shape (m, 3)
output s: np.ndarray of shape (m, 3) where s[i, j] = g(z[i, j])
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# Insert your code here ~ 1-6 line
s = 1.0 / (1.0 + np.exp(-z))
return s
def xent(y_true, y_pred):
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input y_true: np.ndarray of shape (m,)
input y_pred: np.ndarray of shape (m,)
output J: float
\mathbf{I} \cdot \mathbf{I} \cdot \mathbf{I}
# Insert your code here ~ 1-6 lines
m = y_true.shape[0]
a = y_true*np.log(y_pred) # dim (m,)
b = (1-y_true)*np.log(1-y_pred) # dim (m,)
c = -a - b \# dim (m,)
J = np.sum(c) / m
return J
def predict_lr(x, theta):
# Insert your code here ~ 1-3 lines
h = sigmoid(np.dot(x,theta)) # dim (m,1)
y pred = np.round(h) # dim (m,1)
return y pred
```