ReadMe for CS-433 Project 1

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We have submitted to you a folder containing three scripts written in python as well as a sub-folder containing the data you have provided us.

Summary of the contents of the files in the folder.

implementations.py

This script contains exclusively functions that have been used throughout the project. The 6 basic methods that implement the machine learning techniques seen in the courses are included. In addition, we have included other sections where the helpers needed to apply these techniques are provided. Among others, there are sections that calculate cost functions, gradients or allow us to import data. Each section is appropriately titled and the methods that make it up are adequately commented and described.

run.py

This script uses the functions present in the previous section in order to load the data as you provided them to us, clean them, prepare the explanatory variables and perform feature augmentation, train the model, make predictions on unseen data and finally, generate a submission file that is the same as the one we used to obtain the best accuracy score on the AICrowd platform.

cross_validation.py

This script, reproduces the procedure we used to select the parameters of our model. Namely, K-fold cross-validation using a grid search algorithm to select the parameters that allow us to obtain the best average accuracy in cross-validation. Please note that we have used a fairly large grid and therefore this script takes time to complete. At the end of it, it generates a text file with the final selected parameters that were used to generate the predictions.

data folder

This folder contains .csv files. This includes, the data that you have provided us with as well as the predictions file that was used on the AICrowd platform.

Description of the functions.

```
implementations.py
```

```
Models 1.) least_squares_GD(y, tx, initial_w, max_iters, gamma, verbose=False):
Least squares with MSE loss and Gradient Descent.
Parameters
_____
y : Vector
   Dependent variable.
tx : Matrix
   Explanatory variables.
initial_w : Vector
    Initial weights.
max_iters : Integer scalar
   Maximum number of iterations.
gamma : Real scalar
   Learning rate.
verbose : Boolean, optional
   Print each step of Gradient Descent or not. The default is False.
Returns
_____
W : Vector
   Final vector of weights.
loss : Real scalar
   Loss given final weights.
2.) least squares SGD(y, tx, initial w, max iters, gamma, batch size=1, verbose=False):
Least squares with MSE loss and Stochasitc Gradient Descent.
Parameters
_____
y : Vector
   Dependent variable.
tx : Matrix
   Explanatory variables.
initial_w : Vector
   Initial weights.
max_iters : Integer scalar
   Maximum number of iterations.
gamma : Real scalar
   Learning rate.
batch_size : Integer sclar, optional
   Size of the batch. The default is 1.
verbose : Boolean, optional
   Print each step of Gradient Descent or not. The default is False.
```

Returns W : Vector Final vector of weights. loss : Real scalar Loss given final weights. 3.) least_squares(y, tx): Linear regression fit using normal equations. Parameters ----y : Vector Dependent variable. tx : Matrix Explanatory variables. Returns _____ W : Vector Final vector of weights. loss : Real scalar Loss given final weights. 4.) ridge_regression(y, tx, lambda_): Ridge regression fit using normal equations. Parameters ----y : Vector Dependent variable. tx : Matrix Explanatory variables. lambda_: Real scalar Regularization parameter. Returns _____ W : Vector Final vector of weights. loss : Real scalar Loss given final weights. 5.) logistic_regression(y, tx, initial_w, max_iters, gamma, verbose=False): Logistic regression with log loss and Gradient Descent. Parameters _____ y : Vector Dependent variable.

tx : Matrix

initial_w : Vector

Explanatory variables.

Initial weights.

max_iters : Integer scalar

Maximum number of iterations.

gamma : Real scalar Learning rate.

verbose : Boolean, optional

Print each step of Gradient Descent or not. The default is False.

Returns

W : Vector

Final vector of weights.

loss : Real scalar

Loss given final weights.

6.) reg_logistic_regression(y, tx, lambda_, reg, initial_w, max_iters, gamma, verbose=False, early stopping=True, tol=0.0001, patience=5):

Regularized logistic regression with log loss and Gradient Descent with early stopping

Parameters

y : Vector

Dependent variable..

tx : Matrix

Explanatory variables.

lambda_ : Real scalar

Regularization parameter.

reg : Integer scalar

L1 or L2 regularization.

initial_w : Vector

Initial weights.

max_iters : Integer scalar

Maximum number of iterations.

gamma : Real scalar Learning rate.

verbose : Boolean, optional

Print each step of Gradient Descent or not. The default is False.

early_stopping : Boolean, optional

Enable early stopping or not. The default is True.

tol : Real scalar, optional

Minimum amount loss needs to change by. The default is 0.0001.

patience : TYPE, optional

Number of iterations where there must be a decrease in loss by tol. The default is 5.

Returns

W : Vector

Final vector of weights.

loss : Real scalar

Loss given final weights.

Cost functions 1.) mse(y, tx, w):

```
Mean squared error loss function.
Parameters
-----
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
Returns
loss : Real scalar
    MSE loss.
2.) logistic_error(y, tx, w):
Log loss function.
Parameters
_____
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
Returns
loss : Real scalar
    Log loss.
3.) reg_logistic_error(y, tx, w, lambda_, reg):
Log loss function with regularization term.
Parameters
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
lambda_ : Real scalar
    Regularization parameter
reg : Integer scalar
    L1 or L2 regularization
Returns
-----
loss : Real scalar
    Log loss with regularization term.
Gradients 1.) mse_grad(y, tx, w):
```

Compute gradient for MSE loss.

```
-----
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
Returns
_____
gradient : Real scalar
    MSE gradient.
2.) logistic_grad(y, tx, w):
Compute gradient for log loss.
Parameters
-----
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
Returns
gradient : Real scalar
    Log loss gradient.
3.) reg_logistic_grad(y, tx, w, lambda_, reg):
Compute gradient for log loss with regularization term.
Parameters
_____
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
lambda_ : Real scalar
    Regularization parameter
reg : Integer scalar
    {\tt L1} or {\tt L2} regularization
Returns
gradient : Real scalar
    Log loss with regularization term gradient.
Activation functions 1.) sigmoid(x):
Compute sigmoid function
```

Parameters

```
Parameters
_____
x : Vector, Matrix or scalar
    Explanatory variables.
Returns
-----
sigmoid : Vector, Matrix or scalar
    Sigmoid function applied to input.
Helpers 1.) batch_iter(y, tx, batch_size, num_batches=1, shuffle=True):
Generate a minibatch iterator for a dataset.
Parameters
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
batch_size : Integer scalar
    Size of the batch.
num_batches : Integer scalar, optional
    Number of batches. The default is 1.
shuffle : Boolean, optional
    Shuffle data or not. The default is True.
Yields
_____
y : Vector
    Mini-batch of y
tx : Matrix
    Mini-batch of tx
2.) import_data(path="data/"):
Import csv files of train and test data. They must be in the same folder.
Parameters
_____
path : String, optional
    Directory of the files The default is "data/".
Returns
train : Matrix
    Training set.
test : Matrix
    Testing set.
col_names : Vector
    Column names.
3.) create_csv_submission(ids, y_pred, name):
Creates an output file in csv format for submission to Alcrowd
```

Parameters

```
ids : Vector
   Event ids associated with each prediction.
y_pred : Vector
    Predicted class labels.
name : String
    String name of .csv output file to be created.
4.) standardize_numpy(x, mean=None, std=None):
Standardize the original data set.
Parameters
-----
x : Matrix
    Data to standardize.
mean : Vector, optional
    Previously computed mean. The default is None.
std : Vector, optional
    Previously computed standard deviation. The default is None.
Prepare features 1.) split_X_y(train, test, cols):
Create tx matrix for train & test + y vector for train.
Parameters
train : Matrix
    Training set.
test : Matrix
    Testing set.
cols : Vector
    Column names.
Returns
_____
tx train : Matrix
    tx for training set.
y_train : Vector
    y for training set.
tx_test : Matrix
    tx for testing set.
2.) build_poly(x, degree):
Polynomial basis functions for each column of x, for j=1 up to j=degree, and single constant term.
Parameters
-----
x : Matrix
    Matrix to apply augmentation on.
degree : Integer scalar
    $j^{th}$ degree for polynomial basis function.
Returns
-----
phi : Matrix
    Augmented dataset.
```

3.) prepare_features(tx_nan, degree, mean_nan=None, mean=None, std=None):

```
Clean and prepare for learning. Mean imputing, missing value indicator, standardize.
Parameters
_____
tx_nan : Matrix
    Explanatory variables.
degree : Integer scalar
    $j^{th}$ degree for polynomial basis function.
mean_nan : Vector, optional
    Compute column means with if necessary, The default is None.
mean : Vector, optional
    Previously computed mean. The default is None.
std : Vector, optional
    Previously computed standard deviation. The default is None.
Returns
_____
tx : Matrix
    Explanatory variables.
mean : Vector
    Mean of columns.
std : Vector
    Standard deviation of columns.
mean_nan : Vector
    Mean for columns with nan.
nan cols : Vector
    Nan indicator columns.
Performance metrics 1.) logistic prediction(tx, w):
Make a prediction with logistic regression model.
Parameters
_____
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
Returns
y_pred : Vector
    Predictions.
2.) regression_prediction(tx, w):
Make a prediction with linear regression model.
Parameters
tx : Matrix
    Explanatory variables.
w : Vector
    Weights.
```

Returns

```
y_pred : Vector
    Predictions.
3.) f1_score(y_targ, y_pred):
Compute the F1 score of a prediction.
Parameters
-----
y_targ : Vector
    Dependent variable.
y_pred : Vector
    Prediction.
Returns
score : Real scalar
    Score.
4.) accuracy(y targ, y pred):
Compute the accuracy of a prediction.
Parameters
y_targ : Vector
    Dependent variable.
y_pred : Vector
    Prediction.
Returns
score : Real scalar
    Score.
cross_validation.py
1.) cross_validation(y_tr, tx_tr, y_te, tx_te, comb, verbose=2):
Train model, compute in-sample and out-of-sample loss and accuracy.
Parameters
y_tr : Vector
    Dependent variable in training set.
tx_tr : Matrix
    Explanatory variables in training set.
y_te : Vector
    Dependent variable in testing set.
tx_te : Matrix
    Explanatory variables in testing set.
comb : Dictionnary
    Combination of parameters.
    Example: {"gamma":0.1, "lambda":0.01, "reg":2}
verbose : Boolean, optional
    Print progress or not. The default is 2.
Returns
```

```
loss_tr : Real scalar
    Training loss.
loss_te : Real scalar
    Testing loss
f1 : Real scalar
    Testing F1 score.
acc : Real scalar
    Testing accuracy.
2.) model_selection(y, tx, k_fold, degree, grid, seed, verbose=2):
Select the best model from all possible combinations of grid.
Parameters
-----
y : Vector
    Dependent variable.
tx : Matrix
    Explanatory variables.
k_fold : Integer scalar
    Number of folds for cross-validation
degree : Integer scalar
    $j^{th}$ degree for polynomial basis function.
grid : Dictionnary
    Set of all possible combinations.
seed : Integer scalar
    Random seed.
verbose : Boolean, optional
    Print progress or not. The default is 2.
Returns
_____
params : Dictionnary
    Best parameters for given grid.
3.) build k indices(y, k fold, seed):
Build k indices for k-fold.
Parameters
_____
y : Vector
    Dependent variable.
k_fold : Integer scalar
    Number of folds for cross-validation.
seed : Integer scalar
    Random seed.
Returns
_____
k_indices : Matrix
    k indices for K-fold.
4.) prepare_split_data(y, tx, degree, k_fold, seed):
```

Split the dataset based on k-fold cross validation and prepare features.

Parameters

y : Vector

Dependent variable.

tx : Matrix

Explanatory variables.

degree : Integer scalar

\$j^{th}\$ degree for polynomial basis function.

k_fold : Integer scalar

Number of folds for cross-validation.

seed : Integer scalar
Random seed.

Returns

y_trs : Vector

Dependent variable for training set.

tx_trs : Matrix

Explanatory variables for training set.

y_tes : Vector

Dependent variable for testing set.

tx_tes : Matrix

Explanatory variables for testing set..