from sklearn import metrics from keras.preprocessing.text import Tokenizer from keras.preprocessing.sequence import pad sequences from keras.layers import * from keras.models import * from keras import initializers, regularizers, constraints, optimizers, layers from keras.initializers import from keras.optimizers import * import keras.backend as K from keras.callbacks import * import tensorflow as tf import os import time import qc import re from unidecode import unidecode In []: train = pd.read csv("../input/train.csv") test = pd.read_csv("../input/test.csv") print("Train shape : ", train.shape) print("Test shape : ", test.shape) In []: | train["question text"] = train["question text"].str.lower() test["question text"] = test["question text"].str.lower() puncts = [',', '.', '"', ':', ')', '(', '-', '!', '?', '|', ';', "'", '\$', '&', '/', '[', ']', '>', '%' , '=', '#', '*', '+', '\\', '•', '~', '@', '£', $1\cdot 1, \quad 1_1, \quad 1\{1, \quad 1\}1, \quad 1@1, \quad 1^{\wedge 1}, \quad 1@1, \quad 1^{\wedge 1}, \quad 1<1, \quad 1\rightarrow 1, \quad 1^{\otimes 1}, \quad 1^{\otimes 1}, \quad 1 \rightarrow 1, \quad 1 \rightarrow 1,$ 'Â', '**'**', '½', 'à', '...', '``', '*', '"', '-', '•', 'â', '▶', '-', '¢', '²', '¬', '░', '¶', '↑', '±', '¿', '▼', '=', '|', '∥', \tilde{A}^{\dagger} , \tilde{A} '・', ') ', '↓', '、', '│', ' (', '≫', ', '♪', '╩', '╚', '³', '・', '╦', '╣', '╔', '╗', '━', '♥', 'ї', |0|, |1|, |1|, |1|, |1|def clean text(x): x = str(x)for punct in puncts: x = x.replace(punct, f' {punct} ') return x train["question text"] = train["question text"].apply(lambda x: clean text(x)) test["question text"] = test["question text"].apply(lambda x: clean text(x)) In []: ## some config values embed size = 300 # how big is each word vector max features = None # how many unique words to use (i.e num rows in embedding vector) maxlen = 72 # max number of words in a question to use #99.99% ## fill up the missing values X = train["question text"].fillna(" na ").values X test = test["question text"].fillna(" na ").values ## Tokenize the sentences tokenizer = Tokenizer(num words=max features, filters='') tokenizer.fit on texts(list(X)) X = tokenizer.texts to sequences(X) X test = tokenizer.texts to sequences(X test) ## Pad the sentences X = pad sequences(X, maxlen=maxlen) X test = pad sequences(X test, maxlen=maxlen) ## Get the target values Y = train['target'].values sub = test[['qid']] In []: del train, test gc.collect() In []: | word index = tokenizer.word index \max features = len(word index)+1 def load glove(word index): EMBEDDING_FILE = '../input/embeddings/glove.840B.300d/glove.840B.300d.txt' def get_coefs(word, *arr): return word, np.asarray(arr, dtype='float32') embeddings_index = dict(get_coefs(*o.split(" ")) for o in open(EMBEDDING_FILE) if o.split(" ")[0] i n word index) all embs = np.stack(embeddings index.values()) emb mean,emb std = all embs.mean(), all embs.std() embed size = all_embs.shape[1] embedding_matrix = np.random.normal(emb_mean, emb_std, (max_features, embed_size)) for word, i in word index.items(): if i >= max_features: continue embedding_vector = embeddings_index.get(word) if embedding vector is not None: embedding matrix[i] = embedding vector return embedding matrix def load fasttext(word index): EMBEDDING FILE = '../input/embeddings/wiki-news-300d-1M/wiki-news-300d-1M.vec' def get coefs(word, *arr): return word, np.asarray(arr, dtype='float32') embeddings_index = dict(get_coefs(*o.split(" ")) for o in open(EMBEDDING_FILE) if len(o)>100 and o. split(" ")[0] in word index) all embs = np.stack(embeddings index.values()) emb mean,emb std = all embs.mean(), all embs.std() embed size = all_embs.shape[1] embedding_matrix = np.random.normal(emb_mean, emb_std, (max_features, embed_size)) for word, i in word_index.items(): if i >= max features: continue embedding vector = embeddings index.get(word) if embedding vector is not None: embedding matrix[i] = embedding vector return embedding_matrix def load para(word index): EMBEDDING_FILE = '../input/embeddings/paragram_300_s1999/paragram_300_s1999.txt' def get_coefs(word, *arr): return word, np.asarray(arr, dtype='float32') embeddings index = dict(get coefs(*o.split(" ")) for o in open(EMBEDDING FILE, encoding="utf8", err ors='ignore') if len(o)>100 and o.split(" ")[0] in word_index) all embs = np.stack(embeddings index.values()) emb mean,emb std = all embs.mean(), all embs.std() embed size = all embs.shape[1] embedding matrix = np.random.normal(emb mean, emb std, (max features, embed size)) for word, i in word index.items(): if i >= max features: continue embedding vector = embeddings index.get(word) if embedding vector is not None: embedding matrix[i] = embedding vector return embedding matrix In []: embedding matrix 1 = load glove(word index) #embedding matrix 2 = load fasttext(word index) embedding matrix 3 = load para(word index) embedding matrix = np.mean((embedding matrix 1, embedding matrix 3), axis=0) del embedding matrix 1, embedding matrix 3 gc.collect() np.shape(embedding matrix) In []: def squash(x, axis=-1): # s squared norm is really small # s squared norm = K.sum(K.square(x), axis, keepdims=True) + K.epsilon() # scale = K.sqrt(s_squared_norm) / (0.5 + s_squared_norm) # return scale * x s squared norm = K.sum(K.square(x), axis, keepdims=True) scale = K.sqrt(s squared norm + K.epsilon()) return x / scale # A Capsule Implement with Pure Keras class Capsule(Layer): def __init__(self, num_capsule, dim_capsule, routings=3, kernel_size=(9, 1), share_weights=True, activation='default', **kwargs): super(Capsule, self).__init__(**kwargs) self.num_capsule = num_capsule self.dim_capsule = dim_capsule self.routings = routings self.kernel size = kernel size self.share_weights = share_weights if activation == 'default': self.activation = squash else: self.activation = Activation(activation) def build(self, input_shape): super(Capsule, self).build(input shape) input_dim_capsule = input_shape[-1] if self.share_weights: self.W = self.add_weight(name='capsule_kernel', shape=(1, input_dim_capsule, self.num_capsule * self.dim_capsule), # shape=self.kernel_size, initializer='glorot_uniform', trainable=True) input_num_capsule = input_shape[-2] self.W = self.add_weight(name='capsule_kernel', shape=(input_num_capsule, input_dim_capsule, self.num_capsule * self.dim_capsule), initializer='glorot_uniform', trainable=True) def call(self, u_vecs): if self.share_weights: u_hat_vecs = K.conv1d(u_vecs, self.W) else: u_hat_vecs = K.local_conv1d(u_vecs, self.W, [1], [1]) batch size = K.shape(u vecs)[0] input_num_capsule = K.shape(u_vecs)[1] u hat vecs = K.reshape(u_hat_vecs, (batch_size, input_num_capsule, self.num_capsule, self.dim_capsule)) u hat vecs = K.permute_dimensions(u_hat_vecs, (0, 2, 1, 3)) # final u_hat_vecs.shape = [None, num_capsule, input_num_capsule, dim_capsule] b = K.zeros_like(u_hat_vecs[:, :, :, 0]) # shape = [None, num_capsule, input_num_capsule] for i in range(self.routings): b = K.permute_dimensions(b, (0, 2, 1)) # shape = [None, input_num_capsule, num_capsule] c = K.softmax(b) $c = K.permute_dimensions(c, (0, 2, 1))$ b = K.permute dimensions(b, (0, 2, 1))outputs = self.activation(tf.keras.backend.batch_dot(c, u_hat_vecs, [2, 2])) if i < self.routings - 1:</pre> b = tf.keras.backend.batch_dot(outputs, u_hat_vecs, [2, 3]) return outputs def compute_output_shape(self, input_shape): return (None, self.num capsule, self.dim capsule) In []: def capsule(): K.clear session() inp = Input(shape=(maxlen,)) x = Embedding(max_features, embed_size, weights=[embedding_matrix], trainable=**False**)(inp) x = SpatialDropout1D(rate=0.2)(x)x = Bidirectional(CuDNNGRU(100, return_sequences=True, kernel_initializer=glorot_normal(seed=12300), recurrent_initializer=ort hogonal(gain=1.0, seed=10000)))(x)

x = Capsule(num_capsule=10, dim_capsule=10, routings=4, share_weights=**True**)(x)

x = Dense(100, activation="relu", kernel_initializer=glorot_normal(seed=12300))(x)

 $fs = (tp - np.cumsum(y_true[args[:-1]])) / np.arange(y_true.shape[0] + tp - 1, tp, -1)$

X_train, X_val, Y_train, Y_val = X[train_index], X[valid_index], Y[train_index], Y[valid_index]

checkpoint = ModelCheckpoint(filepath, monitor='val_loss', verbose=2, save_best_only=True, mode='mi

reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.6, patience=1, min_lr=0.0001, verbose=2)
earlystopping = EarlyStopping(monitor='val_loss', min_delta=0.0001, patience=2, verbose=2, mode='au

model.fit(X_train, Y_train, batch_size=512, epochs=6, validation_data=(X_val, Y_val), verbose=2, ca

return 2 * fs[res_idx], (y_pred[args[res_idx]] + y_pred[args[res_idx + 1]]) / 2

x = Flatten()(x)

return model

In []: | def f1_smart(y_true, y_pred):

bestscore = []

n')

tp = y_true.sum()

model = capsule()

In []: | y_test = y_test.reshape((-1, 1))

llbacks=callbacks,

x = Dropout(0.12)(x)

x = BatchNormalization()(x)

args = np.argsort(y_pred)

res idx = np.argmax(fs)

y test = np.zeros((X test.shape[0],))

callbacks = [checkpoint, reduce_lr]

if i == 0:print(model.summary())

model.load_weights(filepath)

bestscore.append(threshold)

sub['prediction'] = pred_test_y

sub.to csv("submission.csv", index=False)

filepath="weights_best.h5"

x = Dense(1, activation="sigmoid")(x)
model = Model(inputs=inp, outputs=x)

model.compile(loss='binary_crossentropy', optimizer=Adam(),)

In []: kfold = StratifiedKFold(n_splits=5, random_state=10, shuffle=True)

for i, (train_index, valid_index) in enumerate(kfold.split(X, Y)):

y_pred = model.predict([X_val], batch_size=1024, verbose=2)

pred_test_y = (y_test>np.mean(bestscore)).astype(int)

f1, threshold = f1 smart(np.squeeze(Y val), np.squeeze(y pred))

y_test += np.squeeze(model.predict([X_test], batch_size=1024, verbose=2))/5

print('Optimal F1: {:.4f} at threshold: {:.4f}'.format(f1, threshold))

some reference:

Tell me if missed any

import numpy as np
from tqdm import tqdm

In []: import pandas as pd

tqdm.pandas()

https://www.kaggle.com/shujian/single-rnn-with-4-folds-clr by shujian

https://www.kaggle.com/sudalairajkumar/a-look-at-different-embeddings by SRK

https://www.kaggle.com/christofhenkel/how-to-preprocessing-when-using-embeddings by Dieter

https://www.kaggle.com/shujian/mix-of-nn-models-based-on-meta-embedding by shujian

from sklearn.model_selection import train test split, StratifiedKFold