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NaN  \n",
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NaN  \n",
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NaN  \n",
"3    ham    U dun say so early hor... U c already then say...
NaN  \n",
"4    ham    Nah I don't think he goes to usf, he lives aro...
NaN  \n",
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...\\n",
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oni...\\n",
                "2      Free entry in 2 a wkly comp to win FA Cup
fina...\\n",
                "3      U dun say so early hor... U c already then
say...\\n",
                "4      Nah I don't think he goes to usf, he lives
aro...\\n",
                "
                ...
\\n",
                "5567     This is the 2nd time we have tried 2 contact
u...\\n",
                "5568     Will i_ b going to esplanade fr
home?\\n",
                "5569     Pity, * was in mood for that. So...any other
s...\\n",
                "5570     The guy did some bitching but I acted like
i'd...\\n",
                "5571     Rofl. Its true to its
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                "3      ham\\n",
                "4      ham\\n",
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                ... \\n",
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        "    text_length.append(len(word_tokenize(i)))"
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```

        "from keras.models import Model, load_model\n",
        "from keras.layers import LSTM, Activation, Dense, Dropout,
Input, Embedding\n",
        "from keras.optimizers import RMSprop"
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150\n",
        "    layer = Embedding(vocab_length + 1, 50,
input_length=max_seq_len)(inputs) #None, 150, 50\n",
        "    layer = LSTM(64)(layer)    #None, 64\n",
        "    layer = Dense(256,name='FC1')(layer) #None, 256\n",
        "    layer = Activation('relu')(layer) #None, 256\n",
        "    layer = Dropout(0.5)(layer) #None, 256\n",
        "    layer = Dense(1,name='out_layer')(layer) #None, 1\n",
        "    layer = Activation('sigmoid')(layer) #None, 1\n",
        "    model = Model(inputs=inputs,outputs=layer)\n",
        "
model.compile(loss='binary_crossentropy',optimizer=RMSprop(),
metrics=['acc'])\n",
        "    return model\n",
        "\n",
        "model = create_model(vocab_length, max_sequence_length)\n",
        "model.summary()"
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            \"      \" activation (Activation)        (None, 256)               0
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            \"      \" dropout (Dropout)              (None, 256)               0
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                "28/28 [=====] - 2s 74ms/step -  

loss: 0.0803 - acc: 0.9820 - val_loss: 0.0573 - val_acc: 0.9821\n",
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loss: 0.0268 - acc: 0.9924 - val_loss: 0.0419 - val_acc: 0.9865\n",
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loss: 0.0151 - acc: 0.9961 - val_loss: 0.0412 - val_acc: 0.9843\n",
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loss: 0.0083 - acc: 0.9969 - val_loss: 0.0678 - val_acc: 0.9843\n",
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loss: 0.0052 - acc: 0.9983 - val_loss: 0.0690 - val_acc: 0.9854\n",
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    "plt.show()\n",
    "\n",
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        "print('Test set\\n Loss: {:.3f}\\n Accuracy:\n",
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