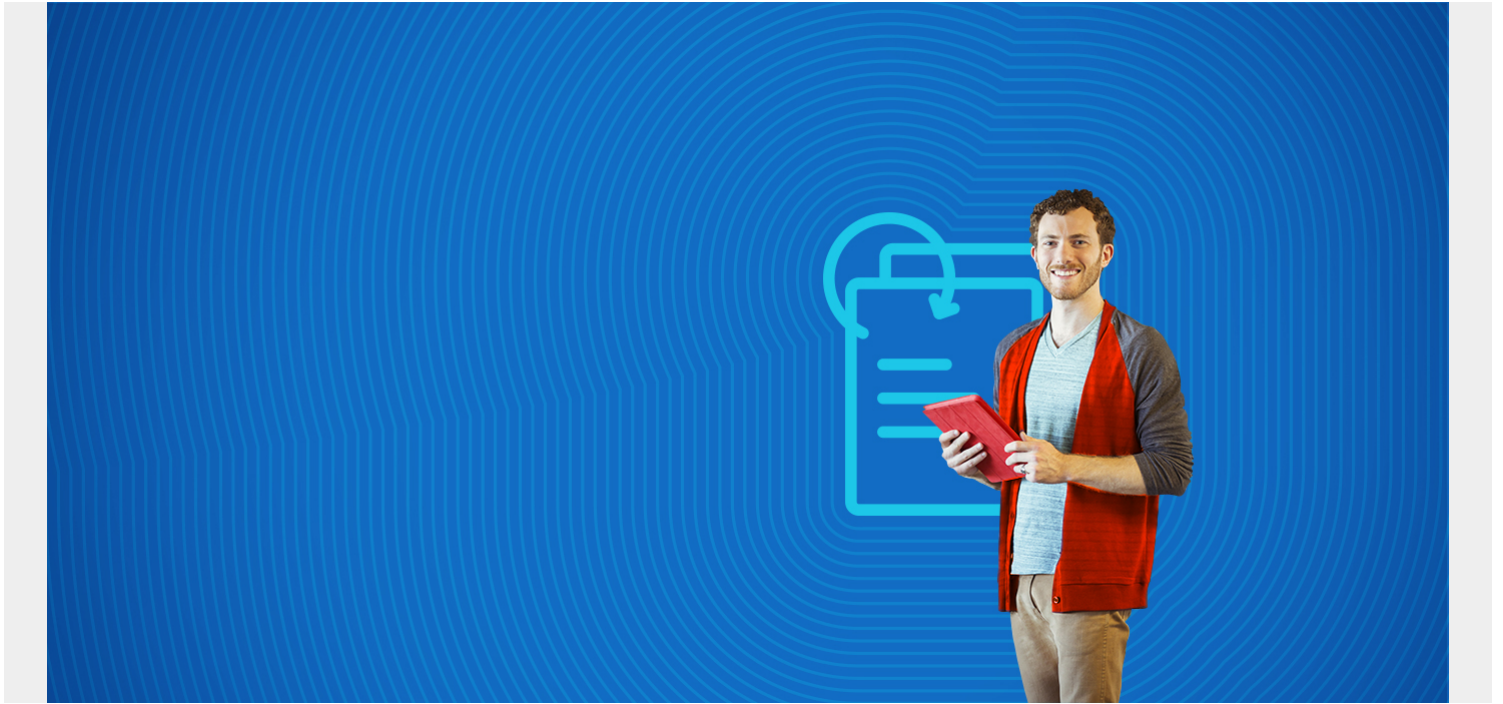


# USING TENSORFLOW TO CREATE A NEURAL NETWORK (WITH EXAMPLES)



When people are trying to learn neural networks with TensorFlow they usually start with the handwriting database. This builds a model that predicts what digit a person has drawn based upon handwriting samples obtained from thousands of persons. To put that into features-labels terms, the combinations of pixels in a grayscale image (white, black, grey) determine what digit is drawn (0, 1, ..., 8, 9).

Here we use other data.

*(This tutorial is part of our [Guide to Machine Learning with TensorFlow & Keras](#). Use the right-hand menu to navigate.)*

## Prerequisites

Before reading this TensorFlow Neural Network tutorial, you should first study these three blog posts:

[Introduction to TensorFlow and Logistic Regression](#)

[What is a Neural Network? Introduction to Neural Networks Part I](#)

[Introduction to Neural Networks Part II](#)

Then you need to install TensorFlow. The easiest way to do that on Ubuntu is to follow [these instructions](#) and use **virtualenv**.

Then install **Python Pandas**, **numpy**, **scikit-learn**, and **SciPy** packages.

# The Las Vegas Strip Hotel Dataset from Trip Advisor

Programmers who are learning to using TensorFlow often start with the iris-data database. That given the combination of pixels that show what type of Iris flower is drawn. But we want to do something original here instead of use the Iris dataset. So we will use the [Las Vegas Strip Data Set](#), cited in the paper "[Moro, S., Rita, P., & Coelho, J. \(2017\). Stripping customers' feedback on hotels through data mining: The case of Las Vegas Strip. Tourism Management Perspectives, 23, 41-52.](#)" and see if we can wrap a neural network around it.

In their paper, the authors wrote a model using the R programming language and used **Support Vector Matrices (SVMs)** as their algorithm. That is a type of non-linear regression problem. It uses the same approach to solving regular LR problems, which is to find a line that reduces the **MSE** (mean square error) to its lowest point to build a predictive model. But SVMs take that up a notch in complexity by working with multiple, nonlinear inputs and finds a plane in n-dimensional space and not line on the XY Cartesian Plane.

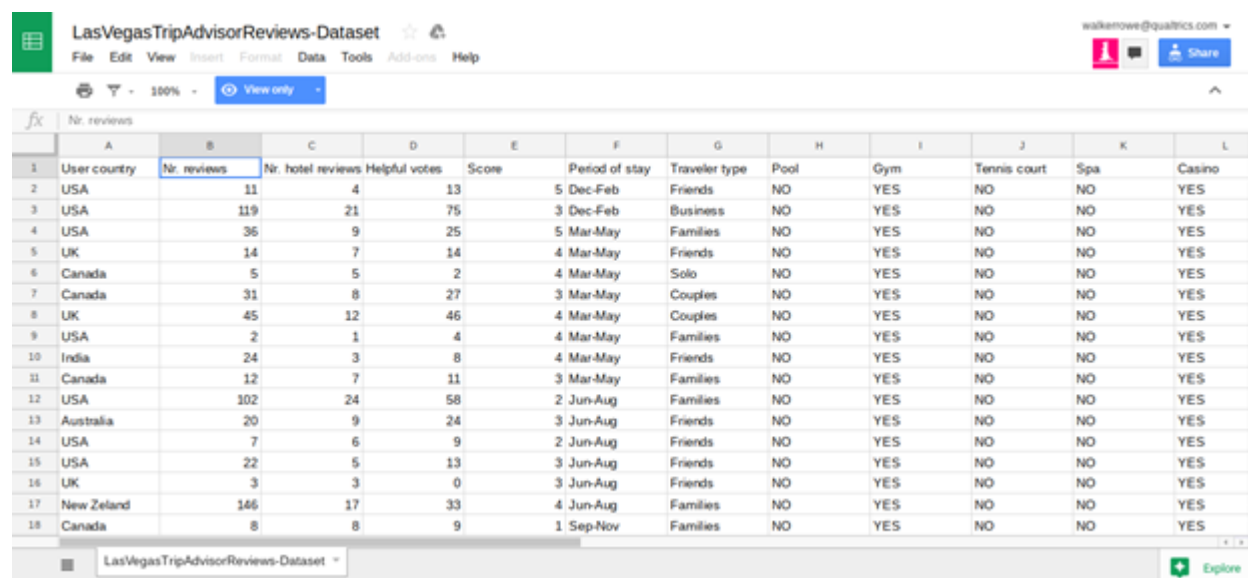
Here we take the same data and but use a neural network instead of SVM. We will present this in 3 blog posts:

1. Put data into numeric format.
2. Train neural network.
3. Make prediction.

The data and code for this tutorial is located [here](#).

## The Data

Click [here to see the](#) data in Google Sheets format. The data is too wide to fit on one screen so we show it below in two screen prints. If you read the paper cited above you can get more details about the data but basically it is TripAdvisor data for 21 Hotels along the Las Vegas Strip. The goal is to build a model that will predict what score an individual is likely to give to which hotel. The score is 1 to 5 and the input are 20 variables described in the spreadsheet below.



The screenshot shows a Google Sheet with the following data:

	A	B	C	D	E	F	G	H	I	J	K	L
1	User country	Nr. reviews	Nr. hotel reviews	Helpful votes	Score	Period of stay	Traveler type	Pool	Gym	Tennis court	Spa	Casino
2	USA	11	4	13	5	Dec-Feb	Friends	NO	YES	NO	NO	YES
3	USA	119	21	75	3	Dec-Feb	Business	NO	YES	NO	NO	YES
4	USA	36	9	25	5	Mar-May	Families	NO	YES	NO	NO	YES
5	UK	14	7	14	4	Mar-May	Friends	NO	YES	NO	NO	YES
6	Canada	5	5	2	4	Mar-May	Solo	NO	YES	NO	NO	YES
7	Canada	31	8	27	3	Mar-May	Couples	NO	YES	NO	NO	YES
8	UK	45	12	46	4	Mar-May	Couples	NO	YES	NO	NO	YES
9	USA	2	1	4	4	Mar-May	Families	NO	YES	NO	NO	YES
10	India	24	3	8	4	Mar-May	Friends	NO	YES	NO	NO	YES
11	Canada	12	7	11	3	Mar-May	Families	NO	YES	NO	NO	YES
12	USA	102	24	58	2	Jun-Aug	Families	NO	YES	NO	NO	YES
13	Australia	20	9	24	3	Jun-Aug	Friends	NO	YES	NO	NO	YES
14	USA	7	6	9	2	Jun-Aug	Friends	NO	YES	NO	NO	YES
15	USA	22	5	13	3	Jun-Aug	Friends	NO	YES	NO	NO	YES
16	UK	3	3	0	3	Jun-Aug	Friends	NO	YES	NO	NO	YES
17	New Zeland	146	17	33	4	Jun-Aug	Families	NO	YES	NO	NO	YES
18	Canada	8	8	9	1	Sep-Nov	Families	NO	YES	NO	NO	YES

	L	M	N	O	P	Q	R	S	T	U	V	W
1	Casino	Free internet	Hotel name	Hotel stars	Nr. rooms	User continent	Member years	Review month	Review weekday			
2	YES	YES	Circus Circus Ho	3	3773	North America	9	January	Thursday			
3	YES	YES	Circus Circus Ho	3	3773	North America	3	January	Friday			
4	YES	YES	Circus Circus Ho	3	3773	North America	2	February	Saturday			
5	YES	YES	Circus Circus Ho	3	3773	Europe	6	February	Friday			
6	YES	YES	Circus Circus Ho	3	3773	North America	7	March	Tuesday			
7	YES	YES	Circus Circus Ho	3	3773	North America	2	March	Tuesday			
8	YES	YES	Circus Circus Ho	3	3773	Europe	4	April	Friday			
9	YES	YES	Circus Circus Ho	3	3773	North America	0	April	Tuesday			
10	YES	YES	Circus Circus Ho	3	3773	Asia	3	May	Saturday			
11	YES	YES	Circus Circus Ho	3	3773	North America	5	May	Tuesday			
12	YES	YES	Circus Circus Ho	3	3773	North America	9	June	Friday			
13	YES	YES	Circus Circus Ho	3	3773	Oceania	4	June	Saturday			
14	YES	YES	Circus Circus Ho	3	3773	North America	1	July	Wednesday			
15	YES	YES	Circus Circus Ho	3	3773	North America	1	July	Thursday			
16	YES	YES	Circus Circus Ho	3	3773	Europe	1	August	Sunday			
17	YES	YES	Circus Circus Ho	3	3773	Oceania	2	August	Saturday			
18	YES	YES	Circus Circus Ho	3	3773	North America	4	September	Wednesday			

The authors of the paper say that certain data—like whether the hotel has a casino, pool, number of stars, or free internet—does not have much bearing on the score given by the hotel guest on TripAdvisor. Rather the factors that most heavily predict the score are the number of reviews the reviewer has written and how long they have been writing reviews. Other factors that influence the score are the day of the week and the month of the year.

## Convert Values to Integers

You can download the code below from [this iPython notebook](#).

First we need to convert all of those values to integers as machine learning uses arrays of numbers as input. We adopt three approaches:

1. If the number is already an integer leave it.
2. If the number is a YES or NO then change it to 1 or 0.
  1. If the element in a string, then use the **ordinal string function** to change each letter to a integer. Then sum those integers.

```
import pandas as pd
```

```
def yesNo(x):
    if x=="YES":
        return 1
    else:
        return 0
```

```
def to0rd(str):
    x=0
    for l in str:
        x += ord(l)
    return int(x)
```

```
cols =
```

```
df = pd.read_csv('/home/walker/TripAdvisor.csv', sep=',', header=0)
```

Here we change every string to an integer. You would have to save the string-integer combination in some data structure so that later you could see which integer equals what string value.

[Here is what our data looks like now.](#)

We've already explained the problem we want to solve, which is to predict how someone might rate one of the Las Vegas hotels on TripAdvisor given how other people have done that. Next we will write the code to build a neural network to do that. In this part we will create the training model. After that we can use our [TensorFlow Neural Network to make predictions](#).

## Prerequisites for building our neural network

- Python 3
- You need to install Tensorflow in Python 3, i.e., **pip3 install --upgrade tensorflow**
- Download this [data](#). This is this [Trip Advisor data](#) converted to integers using [this program](#). It has these column headings. All of these items are features. Score is the label.

```
User country,Nr. reviews,Nr. hotel reviews,Helpful votes,Score,Period of
stay,
Traveler type,Pool,Gym,Tennis court,Spa,Casino,Free internet,Hotel name,
Hotel stars,Nr. rooms>User continent,Member years,Review month,Review weekday
```

Below is the code, which you can copy from [here](#). We explain each section.

Below we put each column name into an array. There are 21 columns. The FIELD\_DEFAULTS are given as 21 integers. We use integers to tell TensorFlow that these are integers and not floats.

```
import tensorflow as tf

feature_names =
FIELD_DEFAULTS = , , , , ,
                , , , , ,
                , , , , ,
                , , , , , ]
```

[Next](#) we want to read the data as a .csv file. Tensorflow provides the **tf.decode\_csv()** method to read one line at a time. We use the dataset **map()** method to call **parse\_line** for each line in the dataset. This creates a **TensorFlow dataset**, which is not a normal Python dataset. It is designed to work with Tensors. If you do not know what a **Tensor** is you can review this.

This routine returns the features as a dictionary and the label as a label. Notice that we delete the Score (**parsed\_line**) from the features since Score is not a feature. It is a label. The **dict(zip())** methods put the key names in the dictionary,

```
def parse_line(line):
    parsed_line = tf.decode_csv(line, FIELD_DEFAULTS)
    label = parsed_line
    del parsed_line
    features = parsed_line
```

```

d = dict(zip(feature_names, features))
print ("dictionary", d, " label = ", label)
return d, label

```

Tensorflow provides the **tf.data.TextLineDataset()** method to read a .csv file into a TensorFlow dataset. **tf.estimator.DNNClassifier.train()** requires that we call some function, in this case **csv\_input\_fn()**, which returns a dataset of features and labels. We use [dataset.shuffle\(\)](#) since that is used when you create neural network.

```

def csv_input_fn(csv_path, batch_size):
    dataset = tf.data.TextLineDataset(csv_path)
    dataset = dataset.map(parse_line)
    dataset = dataset.shuffle(1000).repeat().batch(batch_size)
    return dataset

```

We have to create Tensors for each column in the dataset. We have both categorical data (e.g., 0 and 1) and numbers, e.g., number of reviews.

Categorical data set encode with, e.g., which means there are 47 categories. In other words our same data comes from people from 47 different countries. We can use **df.groupby(df).count()**, for example, to count the unique elements.

```

tf.feature_column.indicator_column(tf.feature_column.categorical_column_with_
identity("Usercountry",47))

```

Numeric data we encode with, for example:

```

Nrreviews = tf.feature_column.numeric_column("Nrreviews")

```

Here is that full section. Notice in the last line we create the array of Tensors in **feature\_columns**.

```

Usercountry =
tf.feature_column.indicator_column(tf.feature_column.categorical_
column_with_identity("Usercountry",47))

Nrreviews = tf.feature_column.numeric_column("Nrreviews")

Nrhotelreviews = tf.feature_column.numeric_column("Nrhotelreviews")

Helpfulvotes = tf.feature_column.numeric_column("Helpfulvotes")

Periodofstay = tf.feature_column.numeric_column("Periodofstay")

Travelertype =
tf.feature_column.indicator_column(tf.feature_column.categorical_column_with_
identity("Travelertype",5))

Pool =
tf.feature_column.indicator_column(tf.feature_column.categorical_
_column_with_identity("Pool",2))

```

```
Gym =
tf.feature_column.indicator_column(tf.feature
_column.categorical_column_with_identity("Gym",2))

Tenniscourt =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Tenniscourt",2))

Spa =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Spa",2))

Casino =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Casino",2))

Freeinternet =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Freeinternet",2))

Hotelname =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Hotelname",24))

Hotelstars =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Hotelstars",5))

Nrrooms =
tf.feature_column.numeric_column("Nrrooms")

Usercontinent =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Usercontinent",6))

Memberyears = tf.feature_column.numeric_column("Memberyears")

Reviewmonth =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Reviewmonth",12))

Reviewweekday =
tf.feature_column.indicator_column(tf.feature_column.categorical
_column_with_identity("Reviewweekday",7))
feature_columns =
```

Here is the [tf.estimator.DNNClassifier](#), where DNN means **Deep Neural Network**. We give it the

feature columns and the directory where it should store the model. We also say there are 5 classes since hotel scores range from 1 to 5. For hidden units we pick . This means the first layer of the neural network has 10 nodes and the next layer has 10. You can read more about how to pick that number by reading, for example, [this StackOverflow article](#). I do not yet know if this is the correct value. We will see when we make predictions in the next post.

```
classifier=tf.estimator.DNNClassifier(  
    feature_columns=feature_columns,  
    hidden_units=,  
    n_classes=5,  
    model_dir="/tmp")
```

```
batch_size = 100
```

Finally we call the **train()** method and give it an inplace (lambda) call to csv\_input\_fn and the path from which we read the csv file.

```
classifier.train(  
    steps=1000,  
    input_fn=lambda : csv_input_fn("/home/walker/tripAdvisorFL.csv", batch_size))
```

[In a separate blog post we will show how to make predictions from this model](#), meaning estimate how a customer might rate a hotel given their characteristics. The hotel could then decide how much effort they might want to expend to make this customer happy or expend no effort at all.