

test1

May 12, 2021

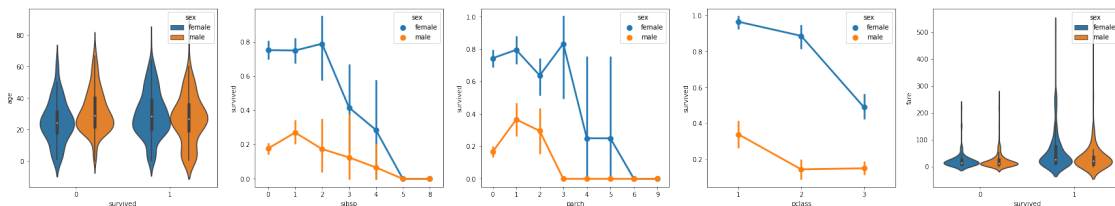
```
[4]: import pandas as pd
import numpy as np
data = pd.read_csv('titanic.csv')
```

```
[5]: data.replace('?', np.nan, inplace=True)
data = data.astype({"age": np.float64, "fare": np.float64})
#data.dtypes
```

```
[6]: import seaborn as sns
import matplotlib.pyplot as plt

fig, axs = plt.subplots(ncols=5, figsize=(30,5))
sns.violinplot(x="survived", y="age", hue="sex", data=data, ax=axs[0])
sns.pointplot(x="sibsp", y="survived", hue="sex", data=data, ax=axs[1])
sns.pointplot(x="parch", y="survived", hue="sex", data=data, ax=axs[2])
sns.pointplot(x="pclass", y="survived", hue="sex", data=data, ax=axs[3])
sns.violinplot(x="survived", y="fare", hue="sex", data=data, ax=axs[4])
```

```
[6]: <AxesSubplot:xlabel='survived', ylabel='fare'>
```



```
[7]: data.replace({'male': 1, 'female': 0}, inplace=True)
```

```
[8]: data.corr().abs()[["survived"]]
```

```
[8]:      survived
pclass    0.312469
survived   1.000000
sex        0.528693
age        0.055513
```

```
sibsp    0.027825
parch    0.082660
fare     0.244265
```

```
[9]: data['relatives'] = data.apply (lambda row: int((row['sibsp'] + row['parch']) > 0), axis=1)
data.corr().abs()[["survived"]]
```

```
[9]:          survived
pclass    0.312469
survived   1.000000
sex        0.528693
age        0.055513
sibsp      0.027825
parch      0.082660
fare       0.244265
relatives  0.201719
```

```
[10]: data = data[['sex', 'pclass', 'age', 'relatives', 'fare', 'survived']].dropna()
```

```
[11]: from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = \
    train_test_split(data[['sex', 'pclass', 'age', 'relatives', 'fare']], data.
    survived, test_size=0.2, random_state=0)
```

```
[12]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(x_train)
X_test = sc.transform(x_test)
```

```
[13]: from sklearn.naive_bayes import GaussianNB
model = GaussianNB()
model.fit(X_train, y_train)
```

```
[13]: GaussianNB()
```

```
[14]: ### Test the model

from sklearn import metrics
predict_test = model.predict(X_test)
print(metrics.accuracy_score(y_test, predict_test))
```

```
0.7464114832535885
```

```
[15]: # Alternatively use Neural Network

from keras.models import Sequential
```

```
from keras.layers import Dense
```

```
model = Sequential()
```

```
[16]: model.add(Dense(5, kernel_initializer = 'uniform', activation = 'relu',  
    ↪input_dim = 5))  
model.add(Dense(5, kernel_initializer = 'uniform', activation = 'relu'))  
model.add(Dense(1, kernel_initializer = 'uniform', activation = 'sigmoid'))
```

```
[17]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 5)	30
dense_1 (Dense)	(None, 5)	30
dense_2 (Dense)	(None, 1)	6

Total params: 66

Trainable params: 66

Non-trainable params: 0

```
[18]: model.compile(optimizer="adam", loss='binary_crossentropy',  
    ↪metrics=['accuracy'])  
model.fit(X_train, y_train, batch_size=32, epochs=50)
```

Epoch 1/50

27/27 [=====] - 1s 22ms/step - loss: 0.6921 - accuracy: 0.5766

Epoch 2/50

27/27 [=====] - 0s 14ms/step - loss: 0.6899 - accuracy: 0.5861

Epoch 3/50

27/27 [=====] - 1s 25ms/step - loss: 0.6860 - accuracy: 0.5861

Epoch 4/50

27/27 [=====] - 0s 13ms/step - loss: 0.6787 - accuracy: 0.5861

Epoch 5/50

27/27 [=====] - 0s 14ms/step - loss: 0.6655 - accuracy: 0.5933

Epoch 6/50

27/27 [=====] - 0s 13ms/step - loss: 0.6461 - accuracy:

0.6699  
Epoch 7/50  
27/27 [=====] - 0s 14ms/step - loss: 0.6214 - accuracy:  
0.7500  
Epoch 8/50  
27/27 [=====] - 0s 17ms/step - loss: 0.5968 - accuracy:  
0.7679  
Epoch 9/50  
27/27 [=====] - 0s 16ms/step - loss: 0.5717 - accuracy:  
0.7739  
Epoch 10/50  
27/27 [=====] - 0s 17ms/step - loss: 0.5498 - accuracy:  
0.7739  
Epoch 11/50  
27/27 [=====] - 1s 42ms/step - loss: 0.5323 - accuracy:  
0.7739  
Epoch 12/50  
27/27 [=====] - 1s 20ms/step - loss: 0.5180 - accuracy:  
0.7703  
Epoch 13/50  
27/27 [=====] - 1s 24ms/step - loss: 0.5069 - accuracy:  
0.7775  
Epoch 14/50  
27/27 [=====] - 0s 15ms/step - loss: 0.4981 - accuracy:  
0.7751  
Epoch 15/50  
27/27 [=====] - 1s 19ms/step - loss: 0.4915 - accuracy:  
0.7799  
Epoch 16/50  
27/27 [=====] - 0s 18ms/step - loss: 0.4870 - accuracy:  
0.7739  
Epoch 17/50  
27/27 [=====] - 1s 36ms/step - loss: 0.4818 - accuracy:  
0.7787  
Epoch 18/50  
27/27 [=====] - 1s 28ms/step - loss: 0.4779 - accuracy:  
0.7775  
Epoch 19/50  
27/27 [=====] - 0s 14ms/step - loss: 0.4744 - accuracy:  
0.7787  
Epoch 20/50  
27/27 [=====] - 1s 32ms/step - loss: 0.4714 - accuracy:  
0.7823  
Epoch 21/50  
27/27 [=====] - 0s 18ms/step - loss: 0.4690 - accuracy:  
0.7811  
Epoch 22/50  
27/27 [=====] - 0s 11ms/step - loss: 0.4668 - accuracy:

0.7811  
 Epoch 23/50  
 27/27 [=====] - 1s 19ms/step - loss: 0.4645 - accuracy:  
 0.7859  
 Epoch 24/50  
 27/27 [=====] - 0s 15ms/step - loss: 0.4628 - accuracy:  
 0.7811  
 Epoch 25/50  
 27/27 [=====] - 0s 11ms/step - loss: 0.4612 - accuracy:  
 0.7799  
 Epoch 26/50  
 27/27 [=====] - 1s 23ms/step - loss: 0.4603 - accuracy:  
 0.7811  
 Epoch 27/50  
 27/27 [=====] - 1s 35ms/step - loss: 0.4588 - accuracy:  
 0.7835  
 Epoch 28/50  
 27/27 [=====] - 1s 23ms/step - loss: 0.4574 - accuracy:  
 0.7883  
 Epoch 29/50  
 27/27 [=====] - 1s 19ms/step - loss: 0.4560 - accuracy:  
 0.7859  
 Epoch 30/50  
 27/27 [=====] - 1s 39ms/step - loss: 0.4550 - accuracy:  
 0.7847  
 Epoch 31/50  
 27/27 [=====] - 0s 11ms/step - loss: 0.4542 - accuracy:  
 0.7835  
 Epoch 32/50  
 27/27 [=====] - 1s 20ms/step - loss: 0.4535 - accuracy:  
 0.7835  
 Epoch 33/50  
 27/27 [=====] - 0s 14ms/step - loss: 0.4524 - accuracy:  
 0.7835  
 Epoch 34/50  
 27/27 [=====] - 0s 11ms/step - loss: 0.4517 - accuracy:  
 0.7883  
 Epoch 35/50  
 27/27 [=====] - 2s 67ms/step - loss: 0.4522 - accuracy:  
 0.7895  
 Epoch 36/50  
 27/27 [=====] - 0s 11ms/step - loss: 0.4502 - accuracy:  
 0.7859  
 Epoch 37/50  
 27/27 [=====] - 1s 21ms/step - loss: 0.4504 - accuracy:  
 0.7847  
 Epoch 38/50  
 27/27 [=====] - 0s 14ms/step - loss: 0.4494 - accuracy:

```

0.7835
Epoch 39/50
27/27 [=====] - 0s 17ms/step - loss: 0.4490 - accuracy:
0.7859
Epoch 40/50
27/27 [=====] - 0s 17ms/step - loss: 0.4488 - accuracy:
0.7835
Epoch 41/50
27/27 [=====] - 0s 17ms/step - loss: 0.4484 - accuracy:
0.7859
Epoch 42/50
27/27 [=====] - 0s 13ms/step - loss: 0.4487 - accuracy:
0.7811
Epoch 43/50
27/27 [=====] - 0s 17ms/step - loss: 0.4479 - accuracy:
0.7823
Epoch 44/50
27/27 [=====] - 1s 33ms/step - loss: 0.4477 - accuracy:
0.7859
Epoch 45/50
27/27 [=====] - 1s 22ms/step - loss: 0.4469 - accuracy:
0.7883
Epoch 46/50
27/27 [=====] - 1s 52ms/step - loss: 0.4471 - accuracy:
0.7895
Epoch 47/50
27/27 [=====] - 1s 24ms/step - loss: 0.4466 - accuracy:
0.7883
Epoch 48/50
27/27 [=====] - 2s 72ms/step - loss: 0.4465 - accuracy:
0.7871
Epoch 49/50
27/27 [=====] - 1s 34ms/step - loss: 0.4461 - accuracy:
0.7883
Epoch 50/50
27/27 [=====] - 0s 17ms/step - loss: 0.4464 - accuracy:
0.7895

```

[18]: <tensorflow.python.keras.callbacks.History at 0x7fc7903b77f0>

```
[19]: y_pred = model.predict_classes(X_test)
      print(metrics.accuracy_score(y_test, y_pred))
```

WARNING:tensorflow:From <ipython-input-19-6007bb8697bc>:1:  
Sequential.predict\_classes (from tensorflow.python.keras.engine.sequential) is  
deprecated and will be removed after 2021-01-01.  
Instructions for updating:  
Please use instead: \* `np.argmax(model.predict(x), axis=-1)`, if your model

```
does multi-class classification (e.g. if it uses a `softmax` last-layer
activation).* `(model.predict(x) > 0.5).astype("int32")`, if your model does
binary classification (e.g. if it uses a `sigmoid` last-layer activation).
0.8038277511961722
```

[ ]: