```
import pandas as pd
pd.plotting.register_matplotlib_converters()
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
print("Setup Complete")
```

Setup Complete

```
# Set up code checking
from learntools.core import binder
binder.bind(globals())
from learntools.data_viz_to_coder.ex5 import *
print("Setup Complete")
```

Setup Complete

```
# Path of the files to read
cancer_filepath = "../input/cancer.csv"

# Fill in the line below to read the file into a variable cancer_data
cancer_data = pd.read_csv(cancer_filepath,index_col="Id")

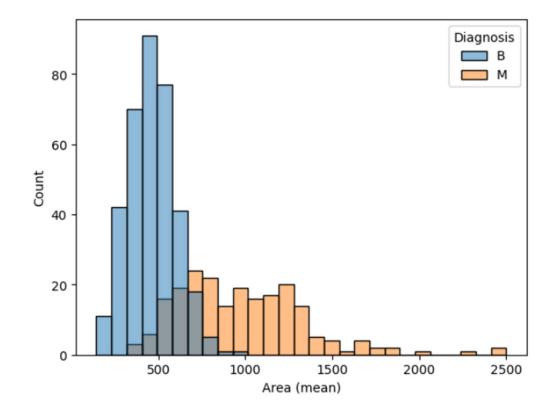
# Run the line below with no changes to check that you've loaded the data correctly
step_1.check()
```

Step2:Print 前五筆資料

	Diagnosis	Radius (mean)	Texture (mean)	Perimeter (mean)	Area (mean)	Smoothness (mean)	Compactness (mean)	Concavity (mean)	Concave points (mean)	Symmetry (mean)	 Radius (worst)	Texture (worst)	Perimeter (worst)	(wor
Id														
8510426	В	13.540	14.36	87.46	566.3	0.09779	0.08129	0.06664	0.047810	0.1885	 15.110	19.26	99.70	71
8510653	В	13.080	15.71	85.63	520.0	0.10750	0.12700	0.04568	0.031100	0.1967	 14.500	20.49	96.09	63
8510824	В	9.504	12.44	60.34	273.9	0.10240	0.06492	0.02956	0.020760	0.1815	 10.230	15.66	65.13	31
854941	В	13.030	18.42	82.61	523.8	0.08983	0.03766	0.02562	0.029230	0.1467	 13.300	22.81	84.46	54
5713702	В	8.196	16.84	51.71	201.9	0.08600	0.05943	0.01588	0.005917	0.1769	 8.964	21.96	57.26	242

Step3:使用下方代碼單元建立兩個直方圖,分別顯示良性和惡性腫瘤的「面積 (平均值)」值的分佈。

```
# Histograms for benign and maligant tumors
sns.histplot(data=cancer_data, x='Area (mean)', hue='Diagnosis')
# Check your answer
step_3.a.check()
```

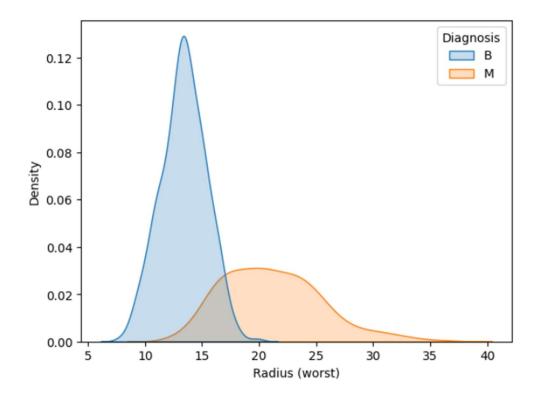


惡性腫瘤的「面積(平均值)」平均數值較高。惡性腫瘤的潛在值範圍較大。

Step4:使用下方代碼單元建立兩個 KDE 圖,分別顯示良性和惡性腫瘤的「半徑(最差)」值的分佈。(為了方便比較,請在下方程式碼單元中建立一個包含兩個 KDE 圖的圖形。)

```
# KDE plots for benign and malignant tumors
sns.kdeplot(data=cancer_data, x='Radius (worst)', hue='Diagnosis', shade=True)

# Check your answer
step_4.a.check()
```



該演算法更有可能將腫瘤歸類為惡性腫瘤。這是因為惡性腫瘤的曲線遠高於良性腫瘤的曲線,其值在 25 左右——而準確率高的演算法很可能會根據數據中的這種模式做出決策。