Game of Thrones - Decision Tree

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• 以DataFrame的形式讀取初始資料集

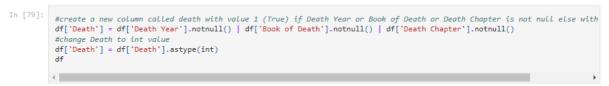
[77]: imp	port pandas as pd												
df df	<pre>df = pd.read_csv('character-deaths.csv') df</pre>												
77]:	Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD
C	Addam Marbrand	Lannister	NaN	NaN	NaN	56.0	1	1	1	1	1	1	0
1	Aegon Frey (Jinglebell)	None	299.0	3.0	51.0	49.0	1	1	0	0	1	0	0
2	Aegon Targaryen	House Targaryen	NaN	NaN	NaN	5.0	1	1	0	0	0	0	1
3	Adrack Humble	House Greyjoy	300.0	5.0	20.0	20.0	1	1	0	0	0	0	1
4	Aemon Costayne	Lannister	NaN	NaN	NaN	NaN	1	1	0	0	1	0	0

912	. Zollo	None	NaN	NaN	NaN	21.0	1	0	0	0	1	0	0
913	Yurkhaz zo Yunzak	None	300.0	5.0	59.0	47.0	1	0	0	0	0	0	1
914	Yezzan Zo Qaggaz	None	300.0	5.0	57.0	25.0	1	1	0	0	0	0	1
915	Torwynd the Tame	Wildling	300.0	5.0	73.0	73.0	1	0	0	0	1	0	0
916 917	Talbert Serry rows × 13 columns	Tyrell	300.0	4.0	29.0	29.0	1	1	0	0	0	1	0

• 檢查各欄位的缺失值狀況

```
In [78]:
         df.isnull().sum()
                               0
Out[78]: Name
         Allegiances
         Death Year
                            612
         Book of Death
                           610
                           618
         Death Chapter
         Book Intro Chapter 12
         Gender
                              0
         Nobility
                               0
         GoT
                               0
         CoK
                               0
         SoS
                               0
         FfC
                               0
         DwD
         dtype: int64
```

• 建立新欄位"Death"以3項欄位值是否為NaN來決定要填入1(死亡)或0(存活)



Out[79]:		Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD	Death
	0	Addam Marbrand	Lannister	NaN	NaN	NaN	56.0	1	1	1	1	1	1	0	0
	1	Aegon Frey (Jinglebell)	None	299.0	3.0	51.0	49.0	1	1	0	0	1	0	0	1
	2	Aegon Targaryen	House Targaryen	NaN	NaN	NaN	5.0	1	1	0	0	0	0	1	0
	3	Adrack Humble	House Greyjoy	300.0	5.0	20.0	20.0	1	1	0	0	0	0	1	1
	4	Aemon Costayne	Lannister	NaN	NaN	NaN	NaN	1	1	0	0	1	0	0	0
	912	Zollo	None	NaN	NaN	NaN	21.0	1	0	0	0	1	0	0	0
	913	Yurkhaz zo Yunzak	None	300.0	5.0	59.0	47.0	1	0	0	0	0	0	1	1
	914	Yezzan Zo Qaggaz	None	300.0	5.0	57.0	25.0	1	1	0	0	0	0	1	1
	915	Torwynd the Tame	Wildling	300.0	5.0	73.0	73.0	1	0	0	0	1	0	0	1
	916	Talbert Serry	Tyrell	300.0	4.0	29.0	29.0	1	1	0	0	0	1	0	1

917 rows × 14 columns

• 刪除作為新欄位"Death"判斷依據的欄位(這些欄位之後不會使用到)

In [80]: #drop Death Year, Book of Death, Death Chapter
df = df.drop(['Death Year', 'Book of Death', 'Death Chapter'], axis=1)
df

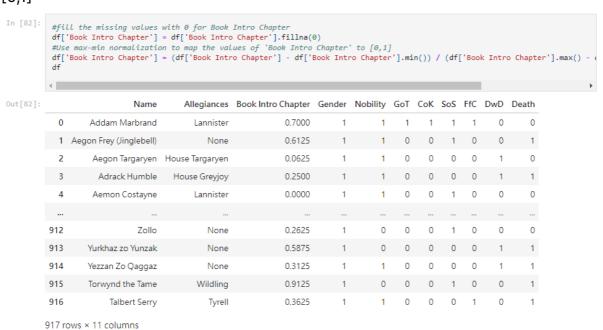
Out[80]:		Name	Allegiances	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD	Death
	0	Addam Marbrand	Lannister	56.0	1	1	1	1	1	1	0	0
	1	Aegon Frey (Jinglebell)	None	49.0	1	1	0	0	1	0	0	1
	2	Aegon Targaryen	House Targaryen	5.0	1	1	0	0	0	0	1	0
	3	Adrack Humble	House Greyjoy	20.0	1	1	0	0	0	0	1	1
	4	Aemon Costayne	Lannister	NaN	1	1	0	0	1	0	0	0
	912	Zollo	None	21.0	1	0	0	0	1	0	0	0
	913	Yurkhaz zo Yunzak	None	47.0	1	0	0	0	0	0	1	1
	914	Yezzan Zo Qaggaz	None	25.0	1	1	0	0	0	0	1	1
	915	Torwynd the Tame	Wildling	73.0	1	0	0	0	1	0	0	1
	916	Talbert Serry	Tyrell	29.0	1	1	0	0	0	1	0	1

917 rows × 11 columns

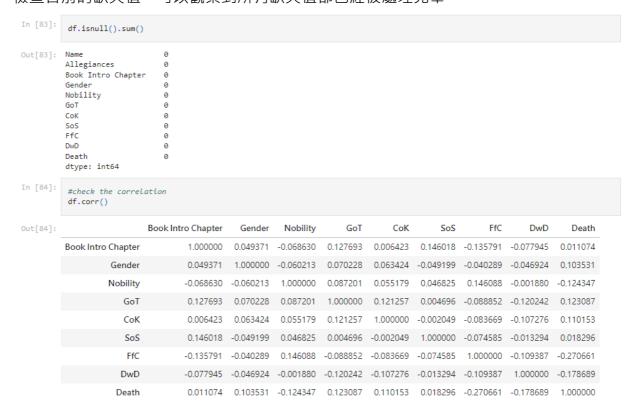
• 檢查目前的欄位缺失值狀況,目前剩"Book Intro Chapter"有缺失值

```
In [81]:
           df.isnull().sum()
Out[81]:
                                    0
                                   0
          Allegiances
          Book Intro Chapter
                                  12
          Gender
                                    0
          Nobility
                                   0
                                   0
          GoT
          CoK
                                    0
          SoS
                                    0
          FfC
                                    0
          DwD
                                    0
          Death
                                    0
          dtype: int64
```

• 用數字0填補缺失值,並且透過Min-Max的轉換,讓"Book Intro Chapter"的欄位值落在 [0,1]



• 檢查目前的缺失值,可以觀察到所有缺失值都已經被處理完畢



• 針對類別資料做編碼的處理,依據原始屬性產生新的二元屬性欄位(0、1)

	Name	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD	Death	 Allegiances_House Tyrell	Allegiances_Lannister	Allegia
0	Addam Marbrand	0.7000	1	1	1	1	1	1	0	0	 0	1	
1	Aegon Frey (Jinglebell)	0.6125	1	1	0	0	1	0	0	1	 0	0	
2	Aegon Targaryen	0.0625	1	1	0	0	0	0	1	0	 0	0	
3	Adrack Humble	0.2500	1	1	0	0	0	0	1	1	 0	0	
4	Aemon Costayne	0.0000	1	1	0	0	1	0	0	0	 0	1	
912	Zollo	0.2625	1	0	0	0	1	0	0	0	 0	0	
913	Yurkhaz zo Yunzak	0.5875	1	0	0	0	0	0	1	1	 0	0	
914	Yezzan Zo Qaggaz	0.3125	1	1	0	0	0	0	1	1	 0	0	
915	Torwynd the Tame	0.9125	1	0	0	0	1	0	0	1	 0	0	
916	Talbert Serry	0.3625	1	1	0	0	0	1	0	1	 0	0	

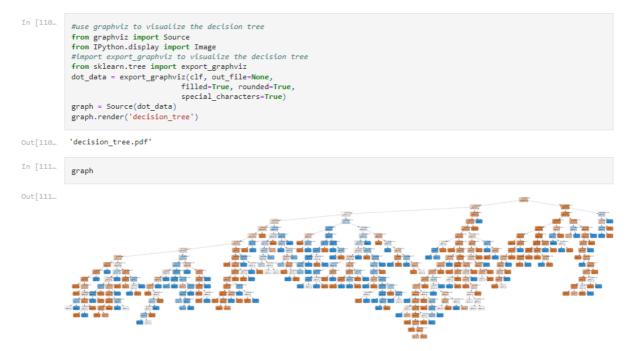
• 準備用來訓練Decision Tree的資料,訓練資料75%測試資料25%,並套用實驗後找到的 指定random_state = 5613(經過20000次的實驗取得,最後產生簡易的confusion

matrix

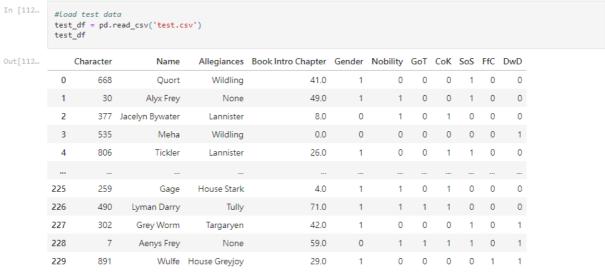
• 印出訓練後的Decision Tree在測試資料集上的precision、recall和accuracy · 並產生更詳細的Decision Tree分類表現報告

```
In [108...
           #show the precision, recall, accuracy
           print("precision: ", precision_score(Y_test, Y_pred))
           print("recall: ", recall_score(Y_test, Y_pred))
           print("accuracy: ", accuracy_score(Y_test, Y_pred))
         precision: 0.7301587301587301
         recall: 0.6216216216216216
         accuracy: 0.8043478260869565
In [109...
           #print classification report
           from sklearn.metrics import classification_report
           print(classification_report(Y_test, Y_pred))
                      precision recall f1-score support
                           0.83 0.89
0.73 0.62
                                               0.86
                                                         156
                    1
                                               0.67
                                                          74
             accuracy
                                               0.80
                                                         230
                                    0.76 0.77
0.80 0.80
                          0.78 0.76
0.80 0.80
                                                         230
           macro avg
                                                         230
         weighted avg
```

• 用graphviz套件繪製訓練出來的Decision Tree的結果



• 以DataFrame載入測試用的資料(之後要提交到Kaggle)

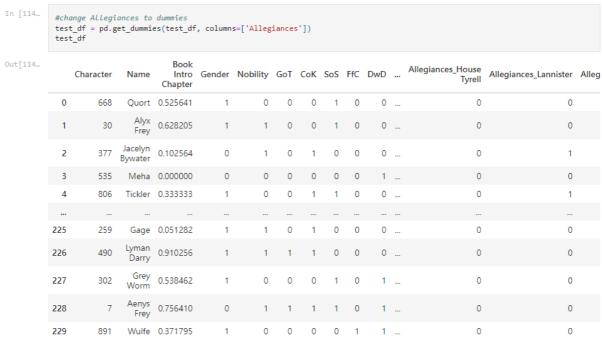


230 rows × 11 columns

• 像前面用一樣的方法處理資料的缺失值和做Min-Max的處理

13	test_ #Use	max-min n df['Book	Intro Chapter'] ormalization to	map the value	ok Intro Chapter']. s of 'Book Intro Ch ook Intro Chapter']	apter' t	o [0,1]	Intro	Chapt	ter'].	.min())) / (t	
13	(Character	Name	Allegiances	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD	
	0	668	Quort	Wildling	0.525641	1	0	0	0	1	0	0	
	1	30	Alyx Frey	None	0.628205	1	1	0	0	1	0	0	
	2	377	Jacelyn Bywater	Lannister	0.102564	0	1	0	1	0	0	0	
	3	535	Meha	Wildling	0.000000	0	0	0	0	0	0	1	
	4	806	Tickler	Lannister	0.333333	1	0	0	1	1	0	0	
	225	259	Gage	House Stark	0.051282	1	1	0	1	0	0	0	
	226	490	Lyman Darry	Tully	0.910256	1	1	1	1	0	0	0	
	227	302	Grey Worm	Targaryen	0.538462	1	0	0	0	1	0	1	
	228	7	Aenys Frey	None	0.756410	0	1	1	1	1	0	1	
	229	891	Wulfe	House Greyjoy	0.371795	1	0	0	0	0	1	1	
2	30 rov	vs × 11 co	lumns										

像前面一樣針對類別資料,按照原始的屬性值,建立許多新的二元值欄位(0、1)



230 rows × 31 columns

刪除預測不需要使用到的資料欄位,並產生final_df

f	<pre>#drop the columns model will not use (Name and Character) final_df = test_df.drop(['Name', 'Character'], axis=1) final_df</pre>													
		Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD	Allegiances_Arryn	Allegiances_Baratheon		Allegiances_House Tyrell	Alleg
	0	0.525641	1	0	0	0	1	0	0	0	0		0	
	1	0.628205	1	1	0	0	1	0	0	0	0		0	
	2	0.102564	0	1	0	1	0	0	0	0	0		0	
	3	0.000000	0	0	0	0	0	0	1	0	0		0	
	4	0.333333	1	0	0	1	1	0	0	0	0		0	
2	25	0.051282	1	1	0	1	0	0	0	0	0		0	
2	26	0.910256	1	1	1	1	0	0	0	0	0		0	
2	27	0.538462	1	0	0	0	1	0	1	0	0		0	
2	28	0.756410	0	1	1	1	1	0	1	0	0		0	
2	29	0.371795	1	0	0	0	0	1	1	0	0		0	

• 將final_df用剛剛訓練出來的Decision Tree進行"Death"的預測,並印出預測結果

```
In [116...
          #predict the Death
          Y_pred = clf.predict(final df)
          Y_pred
Out[116...
         array([1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1,
                1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0,
                0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1,
                1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1,
                0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1,
                0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0,
                1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0,
                0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0,
                0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0,
                1, 0, 0, 0, 0, 0, 1, 0, 0, 0])
```

• 將預測結果補上對應的"Character"欄位,並輸出成可以上傳到Kaggle的CSV檔

```
#create a new dataframe with the original character and death prediction
df_result = pd.DataFrame({'Character': test_df['Character'], 'Death': Y_pred})
df_result
```

Out[117		Character	Death
	0	668	1
	1	30	0
	2	377	1
	3	535	0
	4	806	1
	225	259	0
	226	490	1
	227	302	0
	228	7	0
	229	891	0

230 rows × 2 columns

```
In [118... #save the result
df_result.to_csv('Submission.csv', index=False)
```

• 以下是將預測結果上傳到Kaggle的結果,當中最好的表現為0.89976

