

```
In [58]: 1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5 from sklearn.model_selection import train_test_split
6 from sklearn.linear_model import LogisticRegression
7 from sklearn.metrics import accuracy_score
8 from sklearn.preprocessing import LabelEncoder
9 import warnings
10 warnings.filterwarnings('ignore')
```

```
In [59]: 1 df = pd.read_csv('employee_promotion.csv')
2 df.head()
```

Out[59]:

	employee_id	department	region	education	gender	recruitment_channel	no_of_trainings	aq
0	65438	Sales & Marketing	region_7	Master's & above	f	sourcing	1	:
1	65141	Operations	region_22	Bachelor's	m	other	1	:
2	7513	Sales & Marketing	region_19	Bachelor's	m	sourcing	1	:
3	2542	Sales & Marketing	region_23	Bachelor's	m	other	2	:
4	48945	Technology	region_26	Bachelor's	m	other	1	:

```
In [60]: 1 df.shape
```

Out[60]: (54808, 13)

In [61]: 1 df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 54808 entries, 0 to 54807
Data columns (total 13 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   employee_id           54808 non-null  int64
 1   department            54808 non-null  object
 2   region                54808 non-null  object
 3   education              52399 non-null  object
 4   gender                54808 non-null  object
 5   recruitment_channel    54808 non-null  object
 6   no_of_trainings       54808 non-null  int64
 7   age                   54808 non-null  int64
 8   previous_year_rating  50684 non-null  float64
 9   length_of_service     54808 non-null  int64
10   awards_won            54808 non-null  int64
11   avg_training_score    52248 non-null  float64
12   is_promoted           54808 non-null  int64
dtypes: float64(2), int64(6), object(5)
memory usage: 5.4+ MB
```

In [62]: 1 df.isnull().sum()

```
Out[62]: employee_id           0
department           0
region              0
education            2409
gender              0
recruitment_channel  0
no_of_trainings     0
age                 0
previous_year_rating 4124
length_of_service   0
awards_won          0
avg_training_score   2560
is_promoted         0
dtype: int64
```

In [63]: 1 df['education'].value\_counts()

```
Out[63]: Bachelor's           36669
Master's & above       14925
Below Secondary        805
Name: education, dtype: int64
```

```
In [64]: 1 df['previous_year_rating'].value_counts()
```

```
Out[64]: 3.0    18618
         5.0    11741
         4.0     9877
         1.0     6223
         2.0     4225
         Name: previous_year_rating, dtype: int64
```

```
In [65]: 1 df['avg_training_score'].value_counts()
```

```
Out[65]: 50.0    2716
         49.0    2681
         48.0    2437
         51.0    2347
         60.0    2155
         59.0    2064
         58.0    1898
         52.0    1856
         47.0    1746
         62.0    1450
         82.0    1447
         57.0    1437
         81.0    1357
         53.0    1324
         80.0    1206
         83.0    1198
         84.0    1168
         79.0    1160
         46.0    1136
         85.0    1072
         56.0    1070
         70.0    1055
         63.0    1021
         69.0    1018
         54.0     997
         68.0     935
         78.0     933
         86.0     912
         71.0     898
         55.0     872
         67.0     728
         72.0     725
         64.0     708
         77.0     697
         87.0     655
         65.0     599
         66.0     580
         73.0     523
         76.0     516
         88.0     444
         74.0     433
         75.0     403
         44.0     335
         89.0     301
         90.0     185
         43.0     176
         91.0     117
         92.0      99
         93.0      84
         94.0      65
         42.0      62
         97.0      49
         96.0      48
         95.0      45
```

```

98.0      37
99.0      35
41.0      26
40.0       5
39.0       2
Name: avg_training_score, dtype: int64

```

In [66]: 1 df.describe()

Out[66]:

	employee_id	no_of_trainings	age	previous_year_rating	length_of_service	award
<b>count</b>	54808.000000	54808.000000	54808.000000	50684.000000	54808.000000	54808.0
<b>mean</b>	39195.830627	1.253011	34.803915	3.329256	5.865512	0.0
<b>std</b>	22586.581449	0.609264	7.660169	1.259993	4.265094	0.0
<b>min</b>	1.000000	1.000000	20.000000	1.000000	1.000000	0.0
<b>25%</b>	19669.750000	1.000000	29.000000	3.000000	3.000000	0.0
<b>50%</b>	39225.500000	1.000000	33.000000	3.000000	5.000000	0.0
<b>75%</b>	58730.500000	1.000000	39.000000	4.000000	7.000000	0.0
<b>max</b>	78298.000000	10.000000	60.000000	5.000000	37.000000	1.0

In [67]: 1 df['previous\_year\_rating'].fillna(df['previous\_year\_rating'].mean(),inplace=

In [68]: 1 df['avg\_training\_score'].fillna(df['avg\_training\_score'].mean(),inplace=True

In [69]: 1 print(df['education'].mode())

```

0    Bachelor's
dtype: object

```

In [70]: 1 print(df['education'].mode()[0])

```

Bachelor's

```

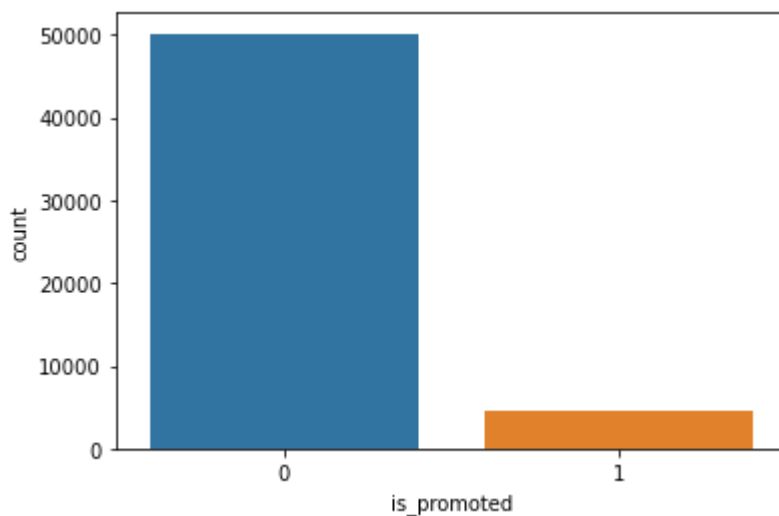
In [71]: 1 df['education'].fillna(df['education'].mode()[0],inplace=True)

```
In [72]: 1 df.isnull().sum()
```

```
Out[72]: employee_id      0
department    0
region        0
education     0
gender        0
recruitment_channel  0
no_of_trainings  0
age           0
previous_year_rating  0
length_of_service  0
awards_won    0
avg_training_score  0
is_promoted   0
dtype: int64
```

```
In [73]: 1 sns.countplot('is_promoted',data = df)
```

```
Out[73]: <AxesSubplot:xlabel='is_promoted', ylabel='count'>
```



```
In [74]: 1 df['is_promoted'].value_counts()
```

```
Out[74]: 0    50140
         1     4668
         Name: is_promoted, dtype: int64
```

```
In [75]: 1 df = df.drop(columns='employee_id',axis = 1)
```

In [76]: 1 df.head()

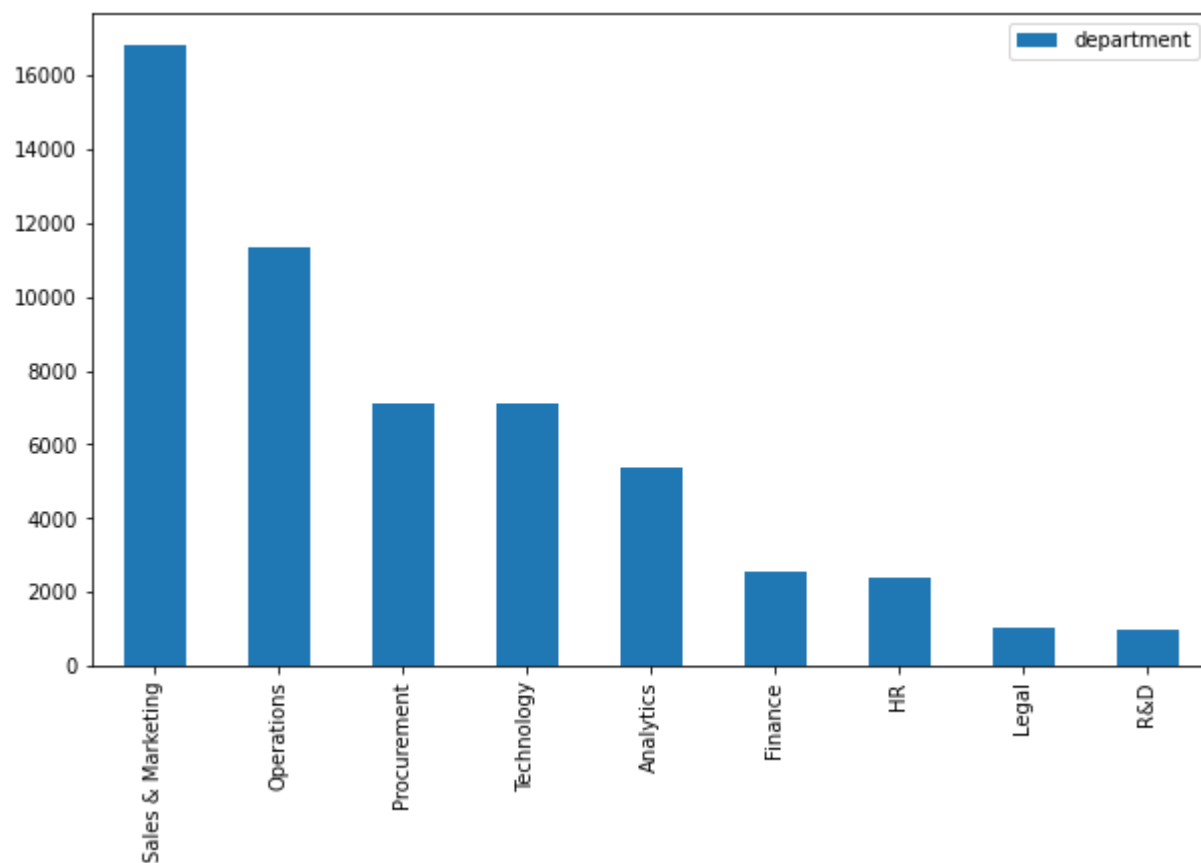
Out[76]:

	department	region	education	gender	recruitment_channel	no_of_trainings	age	previous_y
0	Sales & Marketing	region_7	Master's & above	f	sourcing	1	35	
1	Operations	region_22	Bachelor's	m	other	1	30	
2	Sales & Marketing	region_19	Bachelor's	m	sourcing	1	34	
3	Sales & Marketing	region_23	Bachelor's	m	other	2	39	
4	Technology	region_26	Bachelor's	m	other	1	45	

In [77]: 1 df['department'].value\_counts()

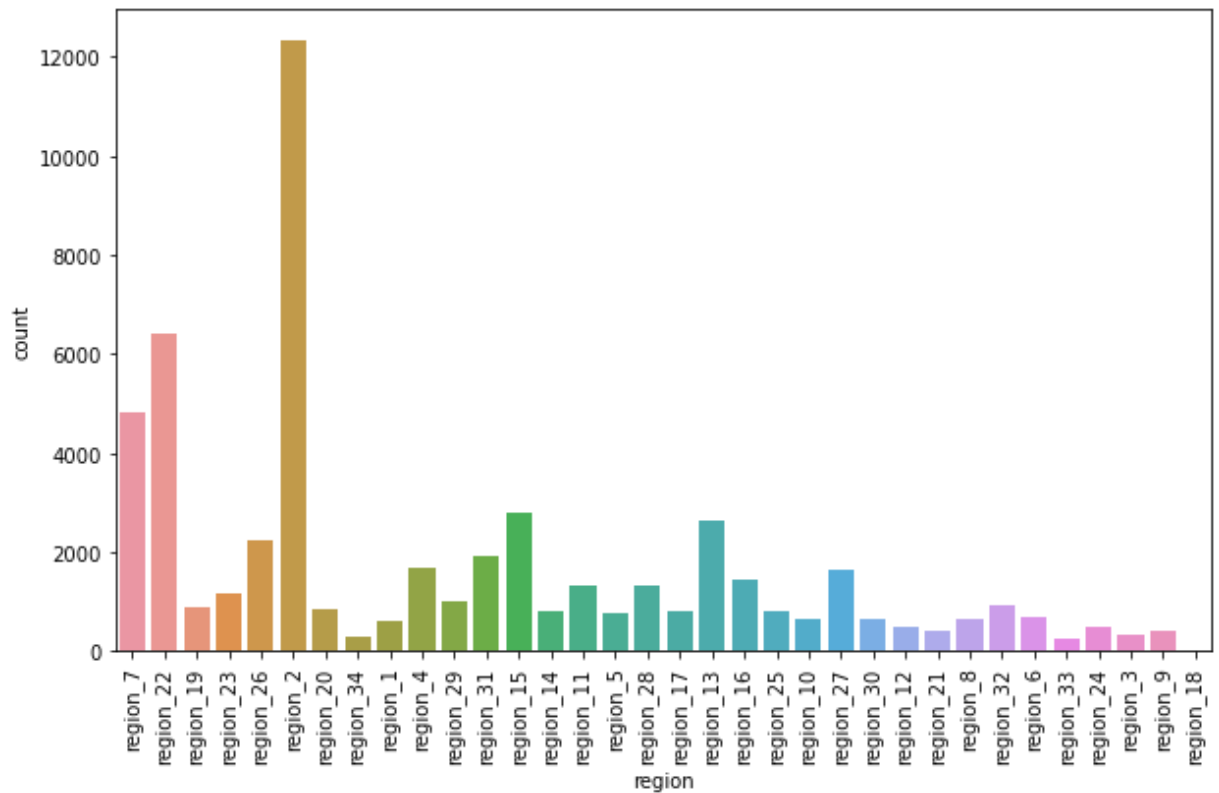
Out[77]: Sales & Marketing 16840  
Operations 11348  
Procurement 7138  
Technology 7138  
Analytics 5352  
Finance 2536  
HR 2418  
Legal 1039  
R&D 999  
Name: department, dtype: int64

```
In [78]: 1 plt.figure(figsize=(10,6))  
2 df['department'].value_counts().plot(kind='bar').legend()  
3 plt.show()
```



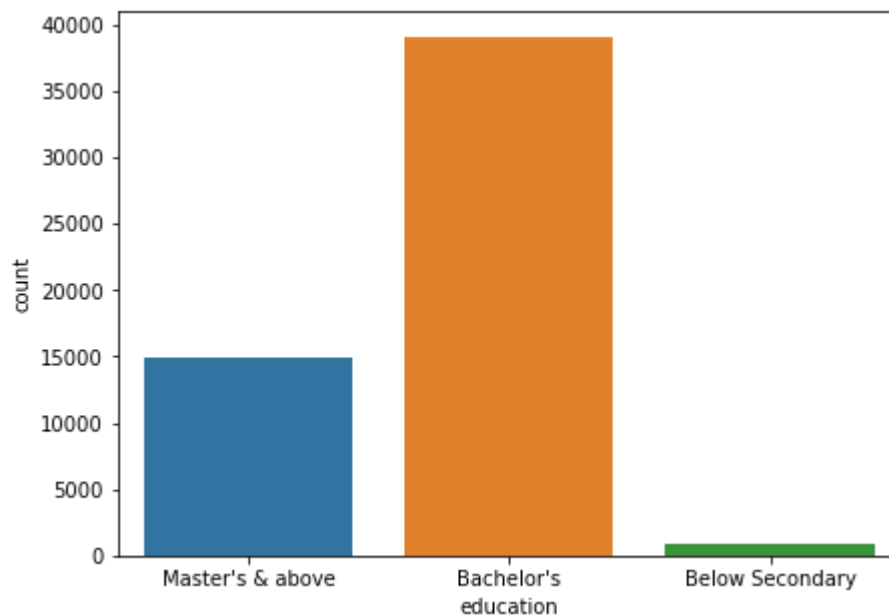


```
In [79]: 1 plt.figure(figsize=(10,6))
2 sns.countplot(data=df,x='region')
3 plt.xticks(rotation=90)
4 plt.show()
```



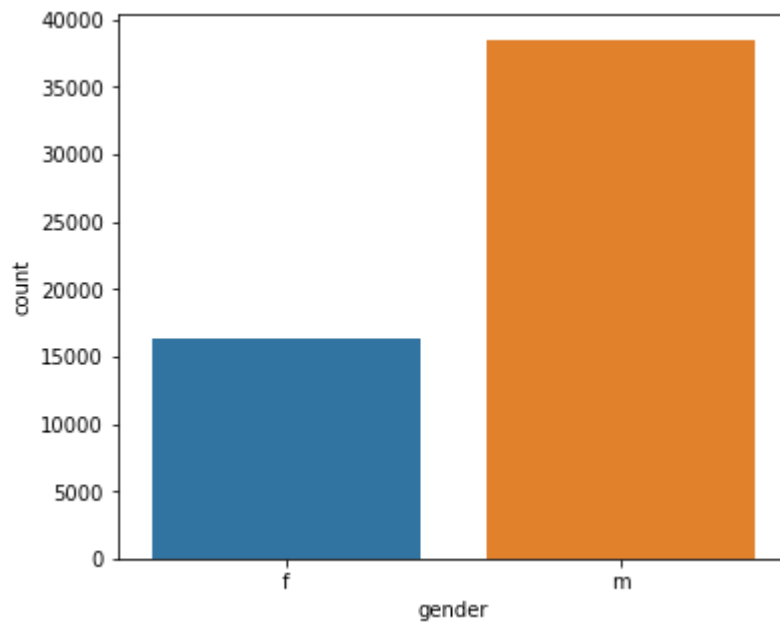
```
In [80]: 1 plt.figure(figsize=(7,5))
2 sns.countplot('education', data=df)
```

Out[80]: <AxesSubplot:xlabel='education', ylabel='count'>



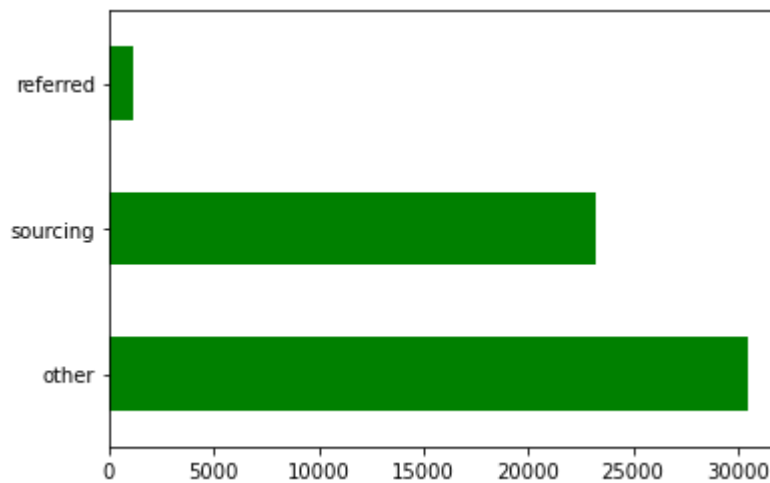
```
In [81]: 1 plt.figure(figsize=(6,5))  
2 sns.countplot('gender', data=df)
```

Out[81]: <AxesSubplot:xlabel='gender', ylabel='count'>

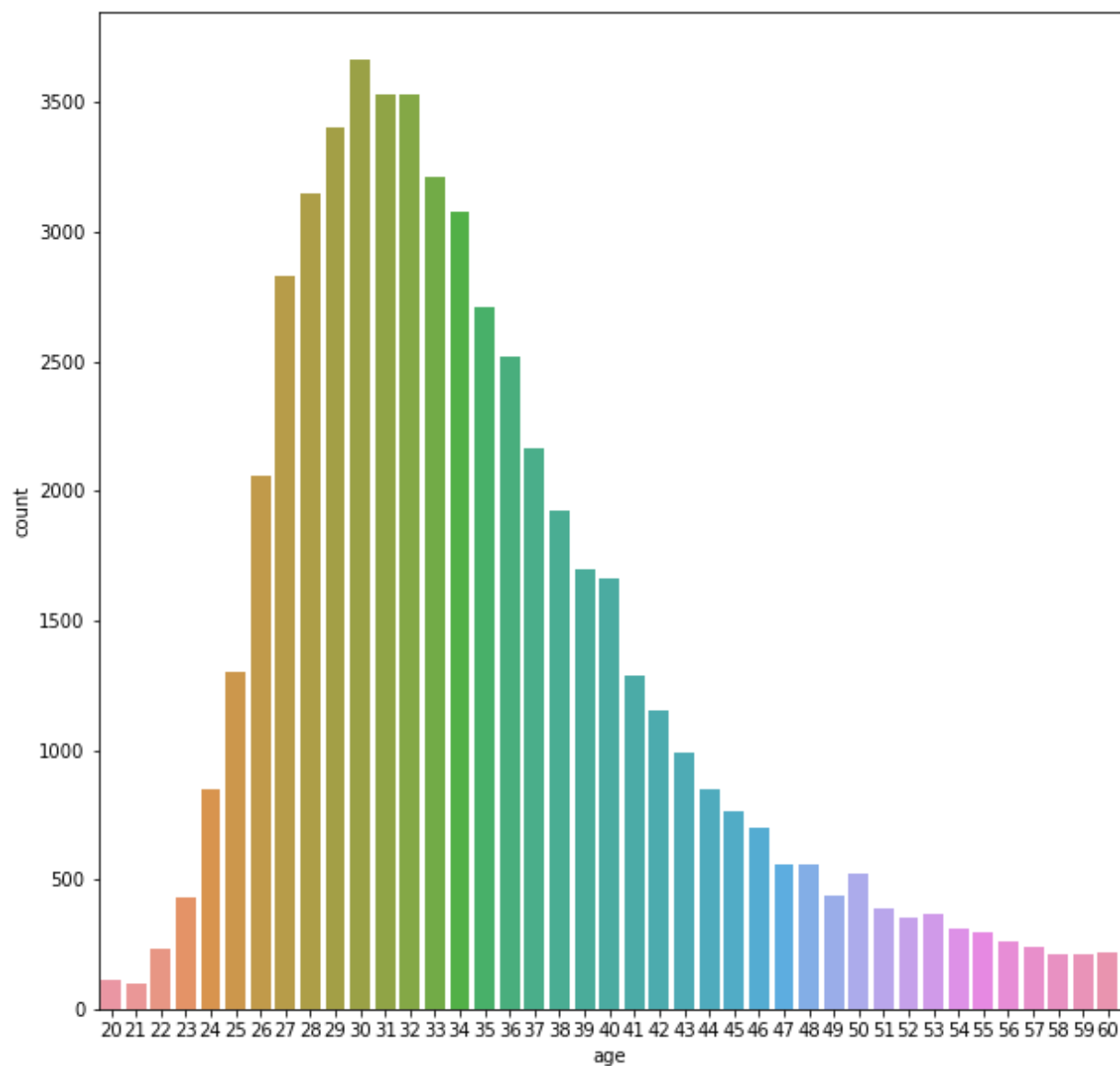


```
In [82]: 1 df['recruitment_channel'].value_counts().plot(kind='barh',color='green')
```

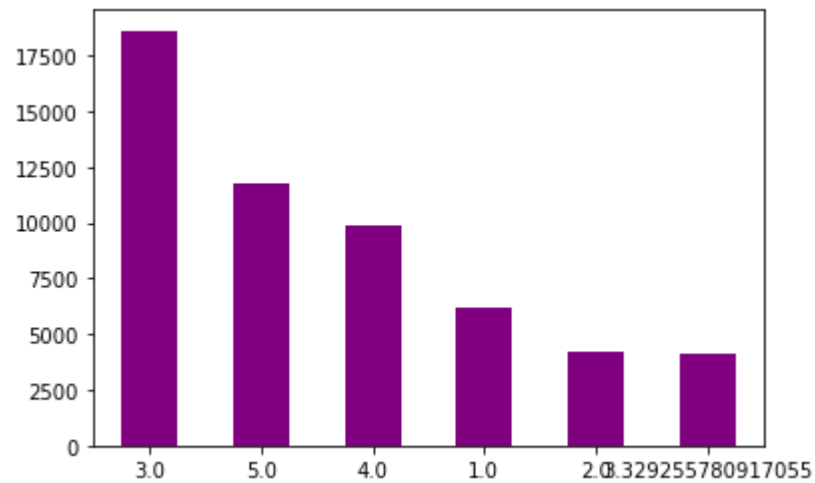
Out[82]: <AxesSubplot:>



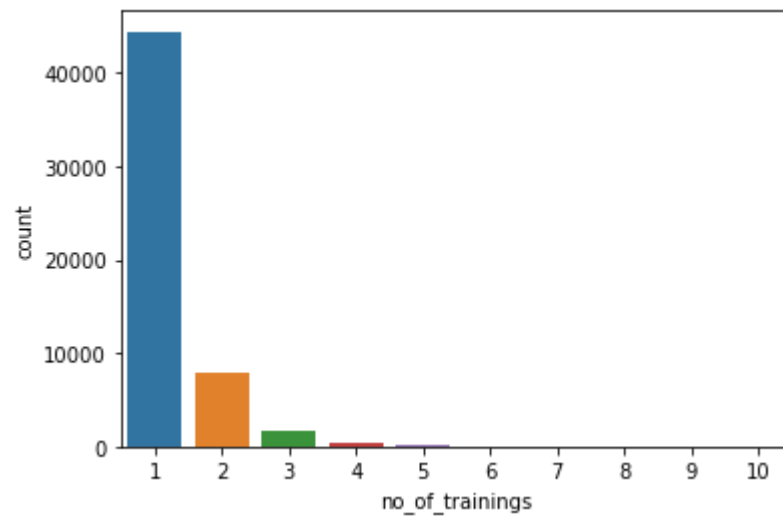
```
In [83]: 1 plt.figure(figsize=(10,10))  
2 sns.countplot(df.age)  
3 plt.show()
```



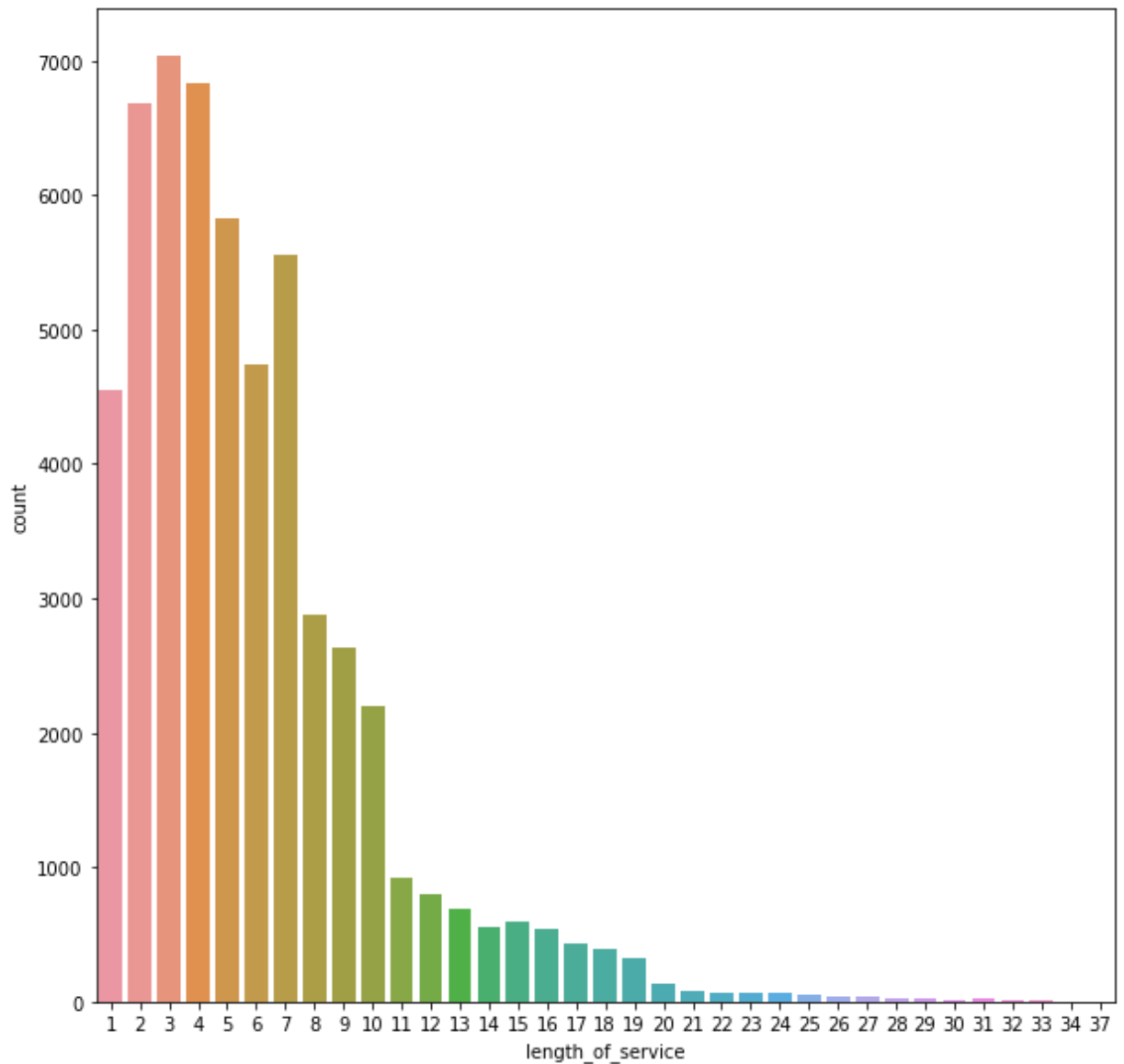
```
In [84]: 1 df['previous_year_rating'].value_counts().plot(kind='bar',color='purple')
2 plt.xticks(rotation=0)
3 plt.show()
```



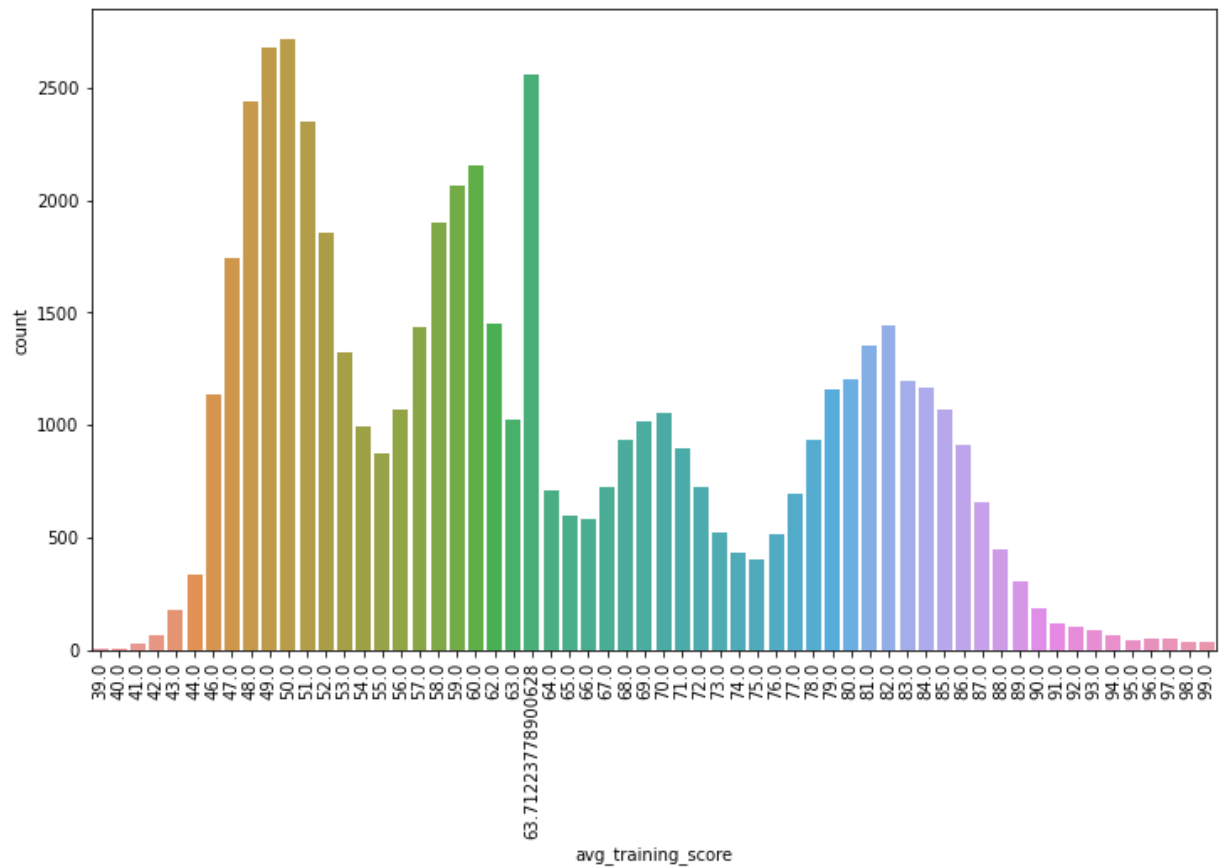
```
In [85]: 1 sns.countplot(df.no_of_trainings)
2 plt.show()
```



```
In [86]: 1 plt.figure(figsize=(10,10))  
2 sns.countplot(df.length_of_service)  
3 plt.show()
```

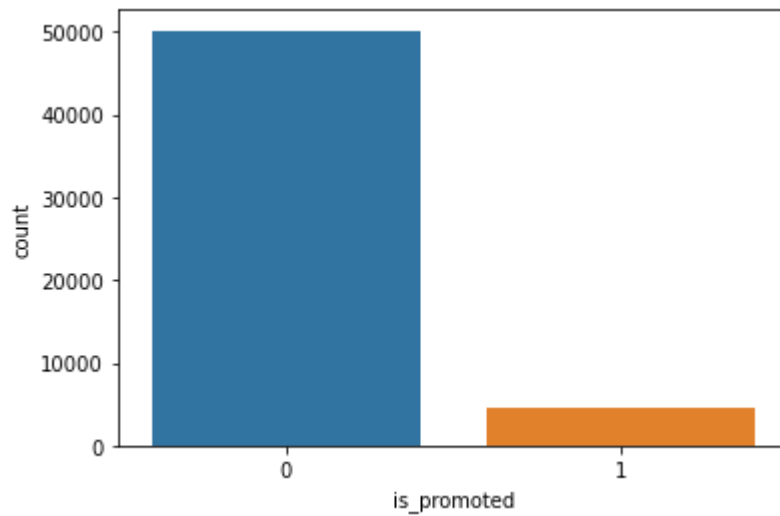


```
In [87]: 1 plt.figure(figsize=(12,7))
2         sns.countplot(df.avg_training_score)
3         plt.xticks(rotation=90)
4         plt.show()
```



```
In [88]: 1 sns.countplot('is_promoted',data = df)
```

```
Out[88]: <AxesSubplot:xlabel='is_promoted', ylabel='count'>
```



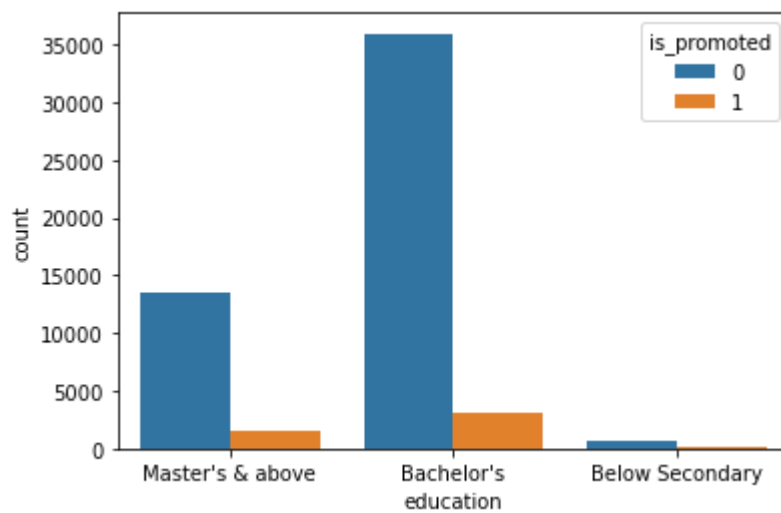
```
In [89]: 1 df.head()
```

```
Out[89]:
```

	department	region	education	gender	recruitment_channel	no_of_trainings	age	previous_y
0	Sales & Marketing	region_7	Master's & above	f	sourcing	1	35	
1	Operations	region_22	Bachelor's	m	other	1	30	
2	Sales & Marketing	region_19	Bachelor's	m	sourcing	1	34	
3	Sales & Marketing	region_23	Bachelor's	m	other	2	39	
4	Technology	region_26	Bachelor's	m	other	1	45	

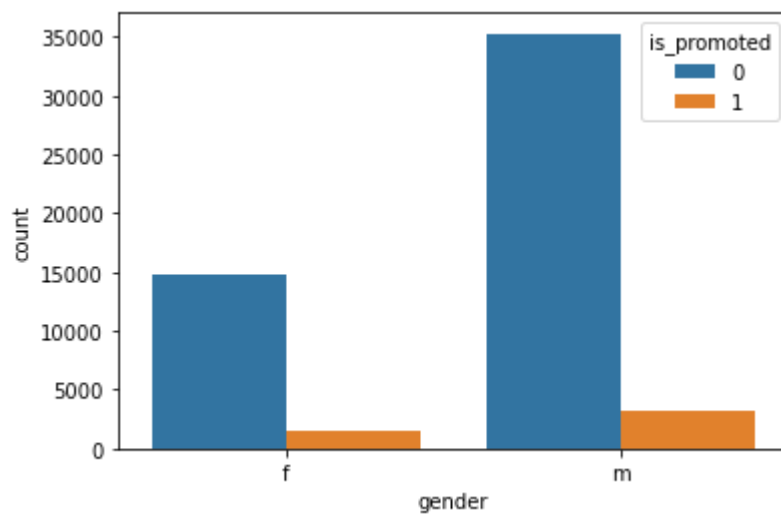
```
In [90]: 1 sns.countplot('education',hue='is_promoted',data=df)
```

```
Out[90]: <AxesSubplot:xlabel='education', ylabel='count'>
```



```
In [91]: 1 sns.countplot('gender',hue='is_promoted',data=df)
```

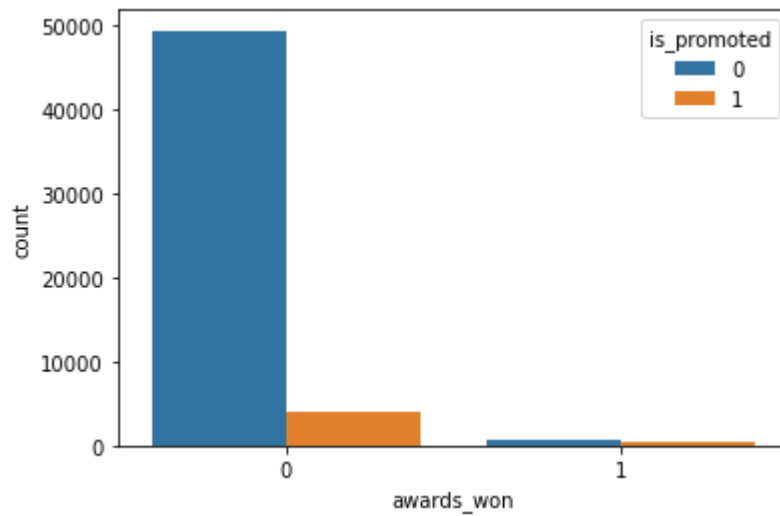
```
Out[91]: <AxesSubplot:xlabel='gender', ylabel='count'>
```





```
In [92]: 1 sns.countplot('awards_won', hue='is_promoted', data=df)
```

```
Out[92]: <AxesSubplot:xlabel='awards_won', ylabel='count'>
```



```
In [93]: 1 le = LabelEncoder()
```

```
In [94]: 1 df['department'] = le.fit_transform(df['department'])
2 df['gender'] = le.fit_transform(df['gender'])
3 df['education'] = le.fit_transform(df['education'].astype(str))
4 df['recruitment_channel'] = le.fit_transform(df['recruitment_channel'])
5 df['region'] = le.fit_transform(df['region'])
```

```
In [95]: 1 df.head()
```

```
Out[95]:
```

	department	region	education	gender	recruitment_channel	no_of_trainings	age	previous_year
0	7	31	2	0	2	1	35	
1	4	14	0	1	0	1	30	
2	7	10	0	1	2	1	34	
3	7	15	0	1	0	2	39	
4	8	18	0	1	0	1	45	

```
In [96]: 1 cols = [ 'previous_year_rating', 'avg_training_score']
2         for col in cols:
3             df[col] = df[col].apply(lambda x: int(x) if x == x else 0)
```

```
In [97]: 1 dc = df.corr()
2         dc
```

Out[97]:

	department	region	education	gender	recruitment_channel	no_of_traini
department	1.000000	-0.020592	0.043441	-0.030108	0.004732	0.014
region	-0.020592	1.000000	-0.003815	0.019730	-0.000920	-0.004
education	0.043441	-0.003815	1.000000	-0.032368	-0.003564	-0.033
gender	-0.030108	0.019730	-0.032368	1.000000	0.006567	0.084
recruitment_channel	0.004732	-0.000920	-0.003564	0.006567	1.000000	-0.010
no_of_trainings	0.014152	-0.004590	-0.033469	0.084501	-0.010405	1.000
age	0.079162	-0.088918	0.339966	-0.016293	-0.011400	-0.081
previous_year_rating	-0.135561	-0.005977	0.027393	-0.024024	0.005731	-0.061
length_of_service	0.059060	-0.058939	0.233666	-0.019675	-0.002887	-0.057
awards_won	-0.002151	0.000307	-0.001409	0.002381	-0.005510	-0.007
avg_training_score	-0.248411	0.023830	0.031798	-0.022761	-0.002650	0.043
is_promoted	0.000130	0.008841	0.029257	-0.011109	0.002229	-0.024

```
In [98]: 1 df.columns
```

Out[98]: Index(['department', 'region', 'education', 'gender', 'recruitment\_channel', 'no\_of\_trainings', 'age', 'previous\_year\_rating', 'length\_of\_service', 'awards\_won', 'avg\_training\_score', 'is\_promoted'], dtype='object')

```
In [99]: 1 X = df[['department', 'region', 'education', 'gender', 'recruitment_channel'
2           'no_of_trainings', 'age', 'previous_year_rating', 'length_of_service'
3           'awards_won', 'avg_training_score']]
4         Y = df[['is_promoted']]
```

In [100]:

1 print(X)

	department	region	education	gender	recruitment_channel	\
0	7	31	2	0		2
1	4	14	0	1		0
2	7	10	0	1		2
3	7	15	0	1		0
4	8	18	0	1		0
...	...	...	...	...		...
54803	8	5	0	1		2
54804	4	19	2	0		0
54805	0	0	0	1		0
54806	7	33	0	1		2
54807	2	14	0	1		0

	no_of_trainings	age	previous_year_rating	length_of_service	\
0	1	35		5	8
1	1	30		5	4
2	1	34		3	7
3	2	39		1	10
4	1	45		3	2
...	...	...		...	...
54803	1	48		3	17
54804	1	37		2	6
54805	1	27		5	3
54806	1	29		1	2
54807	1	27		1	5

	awards_won	avg_training_score
0	0	49
1	0	60
2	0	50
3	0	50
4	0	73
...	...	...
54803	0	78
54804	0	56
54805	0	79
54806	0	63
54807	0	49

[54808 rows x 11 columns]

In [101]: 1 print(Y)

```
      is_promoted
0          0
1          0
2          0
3          0
4          0
...      ...
54803      0
54804      0
54805      0
54806      0
54807      0
```

[54808 rows x 1 columns]

In [102]: 1 X\_train,X\_test,Y\_train,Y\_test = train\_test\_split(X,Y,test\_size=0.3)

In [103]: 1 print(X\_train.shape)  
2 print(Y\_train.shape)  
3 print(X\_test.shape)  
4 print(Y\_test.shape)

```
(38365, 11)
(38365, 1)
(16443, 11)
(16443, 1)
```

In [104]: 1 model = LogisticRegression(solver='liblinear')

In [105]: 1 model.fit(X\_train,Y\_train)

Out[105]: LogisticRegression(solver='liblinear')

In [106]: 1 pred = model.predict(X\_train)

In [107]: 1 training\_data\_accuracy=accuracy\_score(pred,Y\_train)

In [108]: 1 print('Accuracy Score od Training Data: ',training\_data\_accuracy)

Accuracy Score od Training Data: 0.9186498110256744

In [109]: 1 pred1 = model.predict(X\_test)

In [110]: 1 test\_data\_accuracy=accuracy\_score(pred1,Y\_test)

In [111]: 1 `print('Accuracy Score on Test Data: ',test_data_accuracy)`

Accuracy Score on Test Data: 0.9189320683573557

In [112]: 1 `print('Accuracy Score on Training Data: ',training_data_accuracy*100,'%')`  
2 `print('Accuracy Score on Test Data: ',test_data_accuracy*100,'%')`

Accuracy Score on Training Data: 91.86498110256744 %

Accuracy Score on Test Data: 91.89320683573557 %