Brain Stroke Analysis Using Python

I'm thrilled to share the successful completion of a groundbreaking Brain Stroke Analysis project! Here are the key highlights of my work:

Null Value Handling: Identified and meticulously addressed null values within the dataset to ensure impeccable data integrity and accuracy, laying a robust foundation for further analysis.

In-depth Analysis: Conducted a thorough and insightful analysis to decipher patterns and trends related to brain stroke occurrences, providing valuable insights into this critical healthcare domain.

Pata Manipulation Expertise: Employed advanced data manipulation techniques to fill missing values and optimize data quality, enhancing the reliability and usefulness of the dataset.

Python Libraries Mastery: Leveraged powerful Python libraries including Pandas and NumPy for seamless data preprocessing and cleaning, streamlining the analysis process and maximizing efficiency.

ii Visualization Proficiency: Utilized state-of-the-art visualization tools such as Matplotlib and Seaborn to craft visually engaging charts and graphs, effectively communicating complex insights to stakeholders.



Import Library

```
In [1]: import pandas as pd
In [2]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import seaborn as sns
```

C:\Users\Syed Arif\anaconda3\lib\site-packages\scipy__init__.py:146: UserWarnin
g: A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy (d
etected version 1.25.1</pre>

warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"</pre>

Uploading Csv fle

In [3]: df = pd.read_csv(r"C:\Users\Syed Arif\Desktop\healthcare-dataset-stroke-data.csv"

Data Preprocessing

.head()

head is used show to the By default = 5 rows in the dataset

In [4]: df.head()

Out[4]:

	id	gender	age	hypertension	heart_disease	ever_married	work_type	Residence_type	avg_
0	9046	Male	67.0	0	1	Yes	Private	Urban	
1	51676	Female	61.0	0	0	Yes	Self- employed	Rural	
2	31112	Male	80.0	0	1	Yes	Private	Rural	
3	60182	Female	49.0	0	0	Yes	Private	Urban	
4	1665	Female	79.0	1	0	Yes	Self- employed	Rural	
4									•

.tail()

tail is used to show rows by Descending order

```
In [5]: df.tail()
```

Out[5]:

	id	gender	age	hypertension	heart_disease	ever_married	work_type	Residence_type	а
5105	18234	Female	80.0	1	0	Yes	Private	Urban	
5106	44873	Female	81.0	0	0	Yes	Self- employed	Urban	
5107	19723	Female	35.0	0	0	Yes	Self- employed	Rural	
5108	37544	Male	51.0	0	0	Yes	Private	Rural	
5109	44679	Female	44.0	0	0	Yes	Govt_job	Urban	
4									•

.shape

It show the total no of rows & Column in the dataset

```
In [6]: df.shape
Out[6]: (5110, 12)
```

.Columns

It show the no of each Column

.dtypes

This Attribute show the data type of each column

```
In [8]:
        df.dtypes
Out[8]: id
                                int64
        gender
                               object
        age
                              float64
        hypertension
                                int64
        heart_disease
                                int64
        ever_married
                               object
        work_type
                               object
        Residence_type
                               object
        avg_glucose_level
                              float64
                              float64
                               object
        smoking_status
        stroke
                                int64
        dtype: object
```

.unique()

In a column, It show the unique value of specific column.

.nuique()

It will show the total no of unque value from whole data frame

```
In [10]:
         df.nunique()
Out[10]: id
                                5110
          gender
                                   3
                                 104
          age
                                   2
          hypertension
                                   2
          heart_disease
                                   2
          ever_married
                                   5
          work_type
                                   2
          Residence_type
          avg_glucose_level
                                3979
          bmi
                                 418
                                   4
          smoking_status
                                   2
          stroke
          dtype: int64
```

.describe()

It show the Count, mean, median etc

In [11]: df.describe()

Out[11]:

	id	age	hypertension	heart_disease	avg_glucose_level	bmi	
count	5110.000000	5110.000000	5110.000000	5110.000000	5110.000000	4909.000000	5110
mean	36517.829354	43.226614	0.097456	0.054012	106.147677	28.893237	0
std	21161.721625	22.612647	0.296607	0.226063	45.283560	7.854067	0
min	67.000000	0.080000	0.000000	0.000000	55.120000	10.300000	0
25%	17741.250000	25.000000	0.000000	0.000000	77.245000	23.500000	0
50%	36932.000000	45.000000	0.000000	0.000000	91.885000	28.100000	0
75%	54682.000000	61.000000	0.000000	0.000000	114.090000	33.100000	0
max	72940.000000	82.000000	1.000000	1.000000	271.740000	97.600000	1
4							•

.value_counts

It Shows all the unique values with their count

```
In [12]: df["work_type"].value_counts()
```

Out[12]: Private 2925

Self-employed 819 children 687 Govt_job 657 Never_worked 22

Name: work_type, dtype: int64

.isnull()

It shows the how many null values

In [13]: df.isnull()

Out[13]:

	id	gender	age	hypertension	heart_disease	ever_married	work_type	Residence_type	а		
0	False	False	False	False	False	False	False	False			
1	False	False	False	False	False	False	False	False			
2	False	False	False	False	False	False	False	False			
3	False	False	False	False	False	False	False	False			
4	False	False	False	False	False	False	False	False			
5105	False	False	False	False	False	False	False	False			
5106	False	False	False	False	False	False	False	False			
5107	False	False	False	False	False	False	False	False			
5108	False	False	False	False	False	False	False	False			
5109	False	False	False	False	False	False	False	False			
5110 rows × 12 columns											

.info()

To Show Data type of each column

In [14]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5110 entries, 0 to 5109
Data columns (total 12 columns):
Column

#	Column	Non-Null Count	Dtype
0	id	5110 non-null	int64
1	gender	5110 non-null	object
2	age	5110 non-null	float64
3	hypertension	5110 non-null	int64
4	heart_disease	5110 non-null	int64
5	ever_married	5110 non-null	object
6	work_type	5110 non-null	object
7	Residence_type	5110 non-null	object
8	<pre>avg_glucose_level</pre>	5110 non-null	float64
9	bmi	4909 non-null	float64
10	<pre>smoking_status</pre>	5110 non-null	object
11	stroke	5110 non-null	int64
	63 (4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4	/->	

dtypes: float64(3), int64(4), object(5)

memory usage: 479.2+ KB

Is there any Null value present in any Column? Show with heatmap

```
In [15]: df.isnull().sum()
Out[15]: id
                                                   0
              gender
                                                   0
                                                   0
              hypertension
                                                   0
              heart_disease
                                                   0
              ever_married
                                                   0
              work_type
                                                   0
                                                   0
              Residence_type
              avg_glucose_level
                                                   0
                                               201
              bmi
              smoking_status
                                                   0
                                                   0
              stroke
              dtype: int64
In [16]: | sns.heatmap(df.isnull())
Out[16]: <AxesSubplot:>
                                                                                   - 1.0
                244
488
732
976
1220
1464
                                                                                    - 0.8
                1708
1952
                                                                                   - 0.6
                2196
                2684
                2928
3172
                                                                                    0.4
                3416
                3904
                                                                                    0.2
                4148
                4392
4636
4880
                                                                          stroke
                                      hypertension
                                          heart_disease
                                                       Residence_type
                                                            avg_glucose_level
                                              ever_married
                                                   work_type
                                                                     smoking_status
```

Change the Numeric values to categorical

```
In [25]: # Replace 'yes' with 1 and 'no' with 0 in the specified column
df['hypertension'] = df['hypertension'].replace({1: 'yes', 0: 'no'})
```

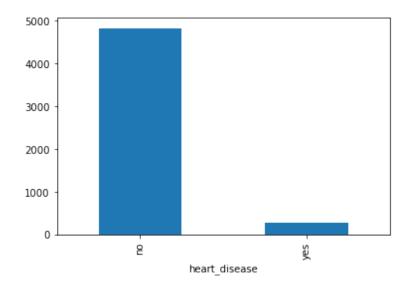
```
In [26]: df.dtypes
Out[26]: id
                                 int64
                               object
         gender
                              float64
         age
         hypertension
                               object
         heart_disease
                                int64
         ever_married
                               object
         work_type
                               object
         Residence_type
                               object
         avg_glucose_level
                              float64
         bmi
                              float64
         smoking_status
                               object
         stroke
                                 int64
         dtype: object
In [27]: # Replace 'yes' with 1 and 'no' with 0 in the specified column
         df['heart_disease'] = df['heart_disease'].replace({1: 'yes', 0: 'no'})
In [28]:
         # Replace 'yes' with 1 and 'no' with 0 in the specified column
         df['stroke'] = df['stroke'].replace({1: 'yes', 0: 'no'})
         Bar Graph
In [29]: | df
Out[29]:
```

	id	gender	age	hypertension	heart_disease	ever_married	work_type	Residence_type	а
0	9046	Male	67.0	no	yes	Yes	Private	Urban	
1	51676	Female	61.0	no	no	Yes	Self- employed	Rural	
2	31112	Male	80.0	no	yes	Yes	Private	Rural	
3	60182	Female	49.0	no	no	Yes	Private	Urban	
4	1665	Female	79.0	yes	no	Yes	Self- employed	Rural	
					•••				
5105	18234	Female	80.0	yes	no	Yes	Private	Urban	
5106	44873	Female	81.0	no	no	Yes	Self- employed	Urban	
5107	19723	Female	35.0	no	no	Yes	Self- employed	Rural	
5108	37544	Male	51.0	no	no	Yes	Private	Rural	
5109	44679	Female	44.0	no	no	Yes	Govt_job	Urban	
5110 r	ows × 1	l2 colum	ns						
4									•

```
In [30]: df.groupby("heart_disease").heart_disease.count()
Out[30]: heart_disease
    no    4834
    yes    276
    Name: heart_disease, dtype: int64
```

```
In [31]: df.groupby("heart_disease").heart_disease.count().plot(kind = "bar")
```

Out[31]: <AxesSubplot:xlabel='heart_disease'>



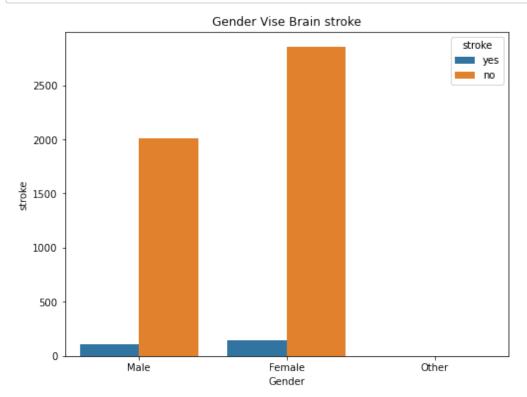
Convert the age column into categorical

```
In [35]: # Apply conditions and replace values in the column
df['age'] = df['age'].apply(lambda x: 'Younger' if x < 18 else ('Young Adult' if</pre>
```

Out[36]:

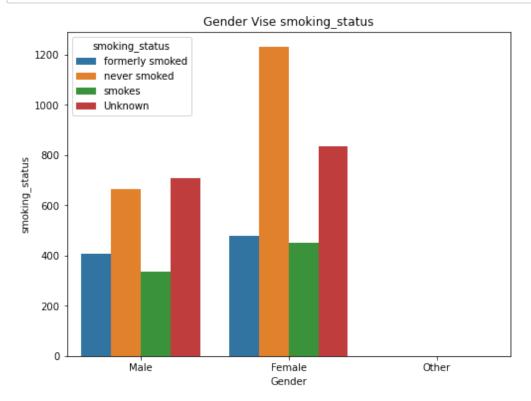
	id	gender	age	hypertension	heart_disease	ever_marrie	d work_type	Residence_type
0	9046	Male	Senior Adult	no	yes	Ye	s Private	Urban
1	51676	Female	Senior Adult	no	no	Ye	s Self- employed	Rural
2	31112	Male	Senior Adult	no	yes	Ye	s Private	Rural
3	60182	Female	Adult	no	no	Ye	s Private	Urban
4	1665	Female	Senior Adult	yes	no	Ye	s Self- employed	Rural
5105	18234	Female	Senior Adult	yes	no	Ye	s Private	Urban
5106	44873	Female	Senior Adult	no	no	Ye	s Self- employed	Urban
5107	19723	Female	Senior Adult	no	no	Υe	s Self- employed	Rural
5108	37544	Male	Senior Adult	no	no	Ye	s Private	Rural
5109	44679	Female	Adult	no	no	Ye	s Govt_job	Urban
5110 r	ows × ′	12 colum	ns					
4						>		

Gender Vise Count of Brain Stroke

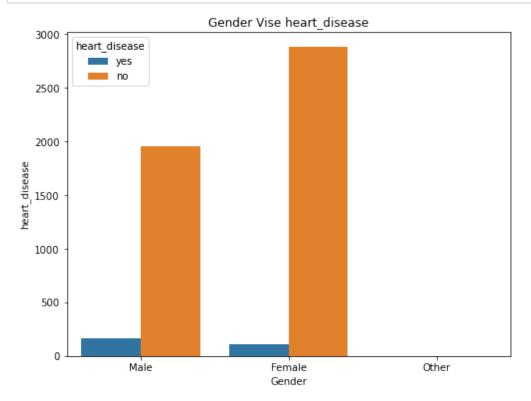


Gender Vise Count of Smoking Status

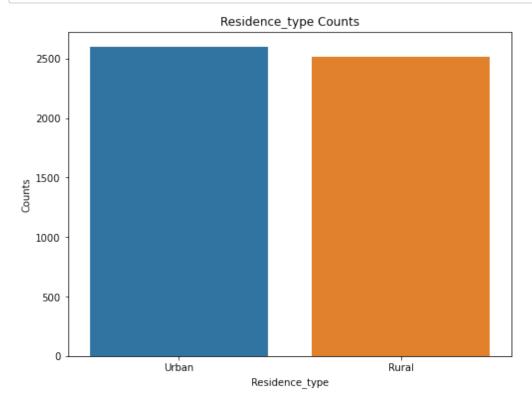
```
In [39]: plt.figure(figsize=(8, 6))
    sns.countplot(data=df, x='gender', hue ="smoking_status")
    plt.xlabel('Gender')
    plt.ylabel('smoking_status')
    plt.title('Gender Vise smoking_status')
    plt.show()
```



Gender Vise Count of Heart Disease



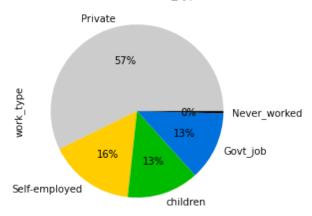
Count of Residency Type



Pie chart for Work Type

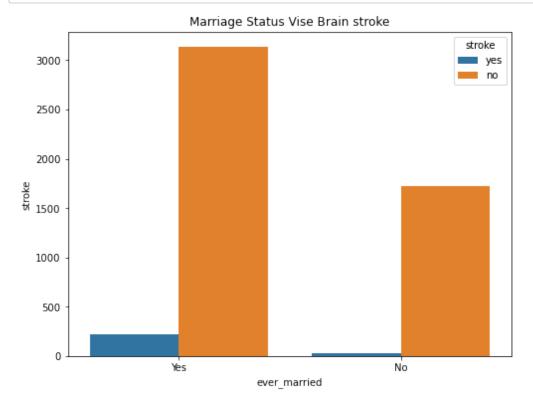
Out[42]: <AxesSubplot:title={'center':'Pie Chart of work_type'}, ylabel='work_type'>



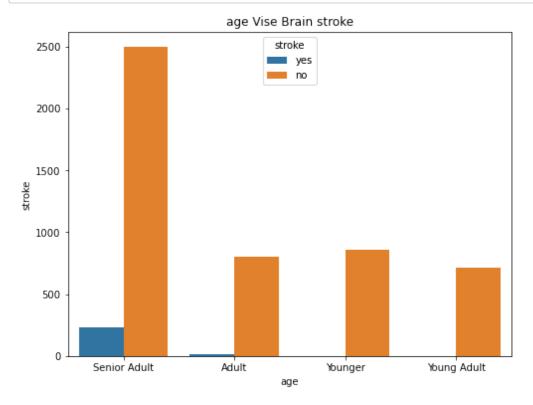


Count of Married Wise Brain Stroke

```
In [43]: plt.figure(figsize=(8, 6))
    sns.countplot(data=df, x='ever_married', hue ="stroke")
    plt.xlabel('ever_married')
    plt.ylabel('stroke')
    plt.title('Marriage Status Vise Brain stroke')
    plt.show()
```



Count of Age Wise Brain Stroke



Count of Age Wise Smoking Status

