

In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

In [2]:

```
data = pd.read_csv("student_alcohol.csv")
```

In [3]:

```
data.head()
```

Out[3]:

	school	sex	age	address	famsize	Pstatus	Medu	Fedu	Mjob	Fjob	...	famrel	freet
0	GP	F	18	U	GT3	A	4	4	at_home	teacher	...	4	
1	GP	F	17	U	GT3	T	1	1	at_home	other	...	5	
2	GP	F	15	U	LE3	T	1	1	at_home	other	...	4	
3	GP	F	15	U	GT3	T	4	2	health	services	...	3	
4	GP	F	16	U	GT3	T	3	3	other	other	...	4	

5 rows × 33 columns

In [4]:

```
data.tail()
```

Out[4]:

	school	sex	age	address	famsize	Pstatus	Medu	Fedu	Mjob	Fjob	...	famrel	freet
390	MS	M	20	U	LE3	A	2	2	services	services	...	5	
391	MS	M	17	U	LE3	T	3	1	services	services	...	2	
392	MS	M	21	R	GT3	T	1	1	other	other	...	5	
393	MS	M	18	R	LE3	T	3	2	services	other	...	4	
394	MS	M	19	U	LE3	T	1	1	other	at_home	...	3	

5 rows × 33 columns

In [5]:



```
data.shape
```

Out[5]:

```
(395, 33)
```

In [6]:



```
data.columns
```

Out[6]:

```
Index(['school', 'sex', 'age', 'address', 'famsize', 'Pstatus', 'Medu', 'Fedu',  
      'Mjob', 'Fjob', 'reason', 'guardian', 'traveltime', 'studytime',  
      'failures', 'schoolsup', 'famsup', 'paid', 'activities', 'nursery',  
      'higher', 'internet', 'romantic', 'famrel', 'freetime', 'goout', 'Dalc',  
      'Walc', 'health', 'absences', 'G1', 'G2', 'G3'],  
      dtype='object')
```

In [7]:



```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 395 entries, 0 to 394
Data columns (total 33 columns):
#   Column          Non-Null Count  Dtype
---  -
0   school          395 non-null    object
1   sex             395 non-null    object
2   age            395 non-null    int64
3   address         395 non-null    object
4   famsize         395 non-null    object
5   Pstatus         395 non-null    object
6   Medu            395 non-null    int64
7   Fedu            395 non-null    int64
8   Mjob            395 non-null    object
9   Fjob            395 non-null    object
10  reason          395 non-null    object
11  guardian        395 non-null    object
12  traveltime      395 non-null    int64
13  studytime       395 non-null    int64
14  failures        395 non-null    int64
15  schoolsup       395 non-null    object
16  famsup          395 non-null    object
17  paid            395 non-null    object
18  activities      395 non-null    object
19  nursery         395 non-null    object
20  higher          395 non-null    object
21  internet        395 non-null    object
22  romantic        395 non-null    object
23  famrel          395 non-null    int64
24  freetime        395 non-null    int64
25  goout           395 non-null    int64
26  Dalc            395 non-null    int64
27  Walc            395 non-null    int64
28  health          395 non-null    int64
29  absences        395 non-null    int64
30  G1              395 non-null    int64
31  G2              395 non-null    int64
32  G3              395 non-null    int64
dtypes: int64(16), object(17)
memory usage: 102.0+ KB
```

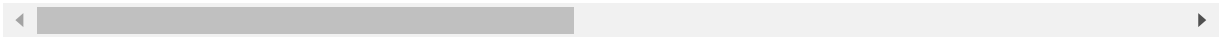
In [8]:



```
data.describe()
```

Out[8]:

	age	Medu	Fedu	traveltime	studytime	failures	famrel	
count	395.000000	395.000000	395.000000	395.000000	395.000000	395.000000	395.000000	3
mean	16.696203	2.749367	2.521519	1.448101	2.035443	0.334177	3.944304	
std	1.276043	1.094735	1.088201	0.697505	0.839240	0.743651	0.896659	
min	15.000000	0.000000	0.000000	1.000000	1.000000	0.000000	1.000000	
25%	16.000000	2.000000	2.000000	1.000000	1.000000	0.000000	4.000000	
50%	17.000000	3.000000	2.000000	1.000000	2.000000	0.000000	4.000000	
75%	18.000000	4.000000	3.000000	2.000000	2.000000	0.000000	5.000000	
max	22.000000	4.000000	4.000000	4.000000	4.000000	3.000000	5.000000	



In [9]:



```
data.isnull().sum()
```

Out[9]:

```
school      0
sex         0
age         0
address     0
famsize     0
Pstatus     0
Medu        0
Fedu        0
Mjob        0
Fjob        0
reason      0
guardian    0
traveltime  0
studytime   0
failures    0
schoolsup   0
famsup      0
paid        0
activities  0
nursery     0
higher      0
internet    0
romantic    0
famrel      0
freetime    0
goout       0
Dalc        0
Walc        0
health      0
absences    0
G1          0
G2          0
G3          0
dtype: int64
```

In [15]:



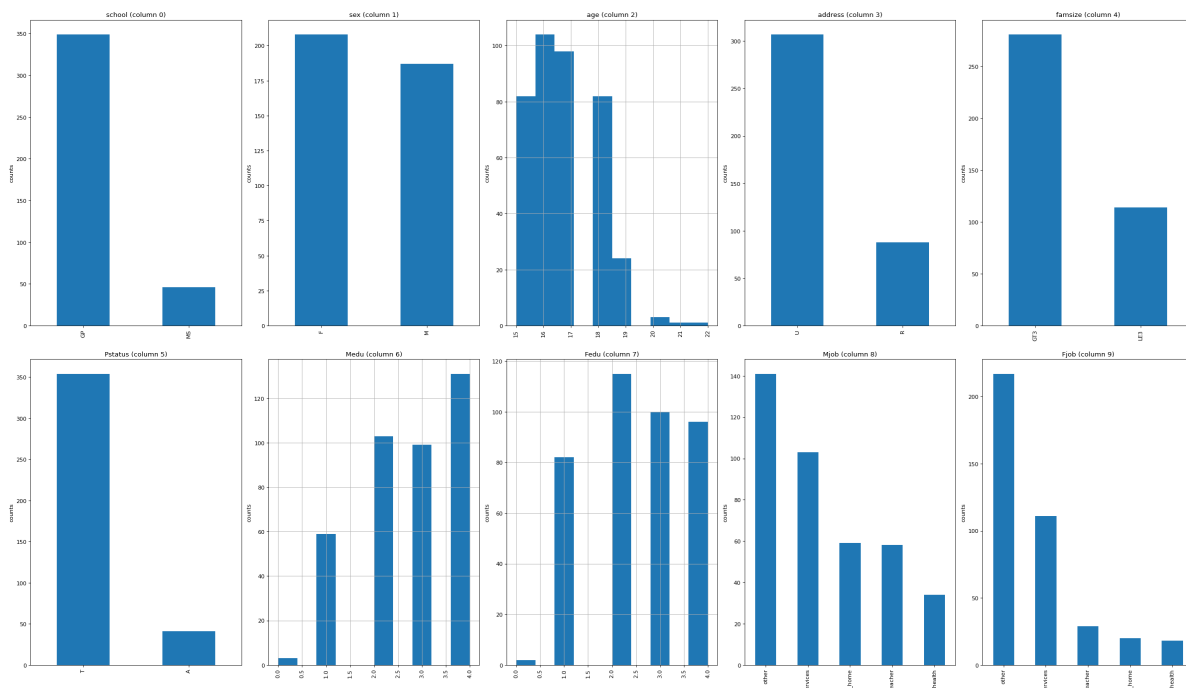
```
import warnings
warnings.filterwarnings('ignore')
```

In [16]:

```
def plotPerColumnDistribution(data, nGraphShown, nGraphPerRow):
    nunique = data.nunique()
    data = data[[col for col in data if nunique[col] > 1 and nunique[col] < 50]]
    nRow, nCol = data.shape
    columnNames = list(data)
    nGraphRow = (nCol + nGraphPerRow - 1) / nGraphPerRow
    plt.figure(num = None, figsize = (6 * nGraphPerRow, 8 * nGraphRow), dpi = 80, facecol = 'white')
    for i in range(min(nCol, nGraphShown)):
        plt.subplot(nGraphRow, nGraphPerRow, i + 1)
        columnDf = data.iloc[:, i]
        if (not np.issubdtype(type(columnDf.iloc[0]), np.number)):
            valueCounts = columnDf.value_counts()
            valueCounts.plot.bar()
        else:
            columnDf.hist()
        plt.ylabel('counts')
        plt.xticks(rotation = 90)
        plt.title(f'{columnNames[i]} (column {i})')
    plt.tight_layout(pad = 1.0, w_pad = 1.0, h_pad = 1.0)
    plt.show()
```

In [17]:

```
plotPerColumnDistribution(data, 10, 5)
```



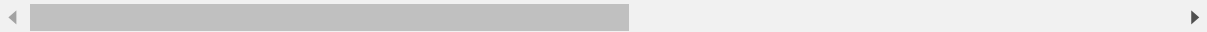
In [22]:



```
data.corr()
```

Out[22]:

	age	Medu	Fedu	traveltime	studytime	failures	famrel	freetime
age	1.000000	-0.163658	-0.163438	0.070641	-0.004140	0.243665	0.053940	0.016434
Medu	-0.163658	1.000000	0.623455	-0.171639	0.064944	-0.236680	-0.003914	0.030891
Fedu	-0.163438	0.623455	1.000000	-0.158194	-0.009175	-0.250408	-0.001370	-0.012846
traveltime	0.070641	-0.171639	-0.158194	1.000000	-0.100909	0.092239	-0.016808	-0.017025
studytime	-0.004140	0.064944	-0.009175	-0.100909	1.000000	-0.173563	0.039731	-0.143198
failures	0.243665	-0.236680	-0.250408	0.092239	-0.173563	1.000000	-0.044337	0.091987
famrel	0.053940	-0.003914	-0.001370	-0.016808	0.039731	-0.044337	1.000000	0.150701
freetime	0.016434	0.030891	-0.012846	-0.017025	-0.143198	0.091987	0.150701	1.000000
goout	0.126964	0.064094	0.043105	0.028540	-0.063904	0.124561	0.064568	0.285011
Dalc	0.131125	0.019834	0.002386	0.138325	-0.196019	0.136047	-0.077594	0.209000
Walc	0.117276	-0.047123	-0.012631	0.134116	-0.253785	0.141962	-0.113397	0.147821
health	-0.062187	-0.046878	0.014742	0.007501	-0.075616	0.065827	0.094056	0.075733
absences	0.175230	0.100285	0.024473	-0.012944	-0.062700	0.063726	-0.044354	-0.058071
G1	-0.064081	0.205341	0.190270	-0.093040	0.160612	-0.354718	0.022168	0.012611
G2	-0.143474	0.215527	0.164893	-0.153198	0.135880	-0.355896	-0.018281	-0.013771
G3	-0.161579	0.217147	0.152457	-0.117142	0.097820	-0.360415	0.051363	0.011301

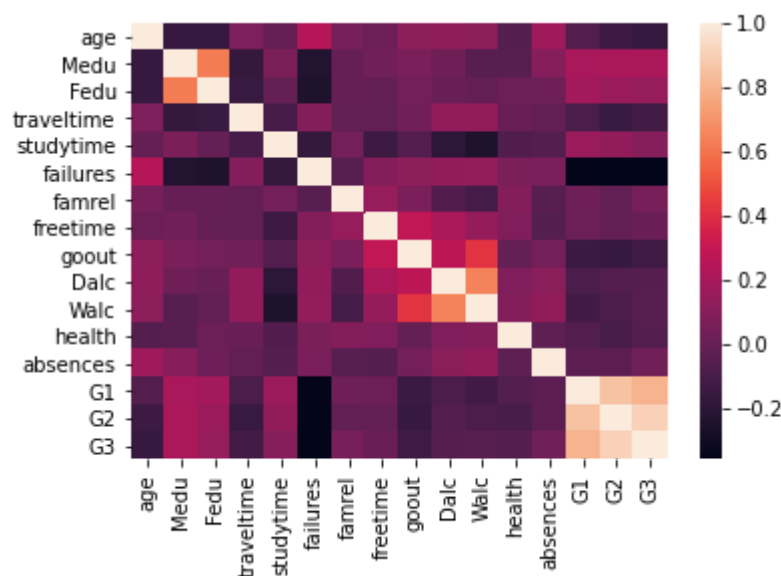


In [23]:

```
sns.heatmap(data.corr())
```

Out[23]:

<AxesSubplot:>

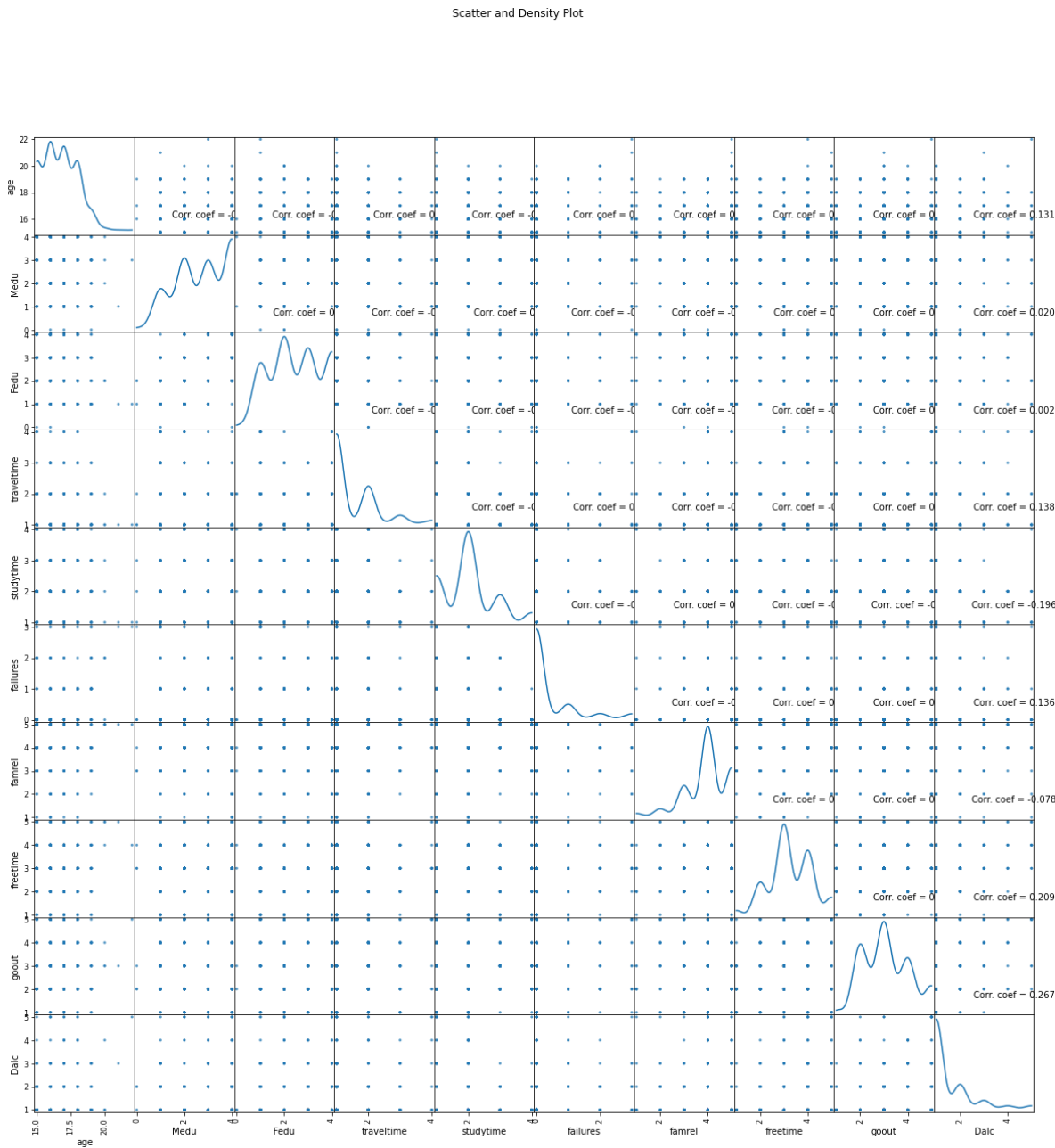


In [24]:

```
def plotScatterMatrix(data, plotSize, textSize):
    data = data.select_dtypes(include=[np.number])
    data = data.dropna('columns')
    data = data[[col for col in data if data[col].nunique() > 1]]
    columnNames = list(data)
    if len(columnNames) > 10:
        columnNames = columnNames[:10]
    data = data[columnNames]
    ax = pd.plotting.scatter_matrix(data, alpha=0.75, figsize=[plotSize, plotSize], diag=True)
    corrs = data.corr().values
    for i, j in zip(*plt.np.triu_indices_from(ax, k = 1)):
        ax[i, j].annotate('Corr. coef = %.3f' % corrs[i, j], (0.8, 0.2), xycoords='axes')
    plt.suptitle('Scatter and Density Plot')
    plt.show()
```

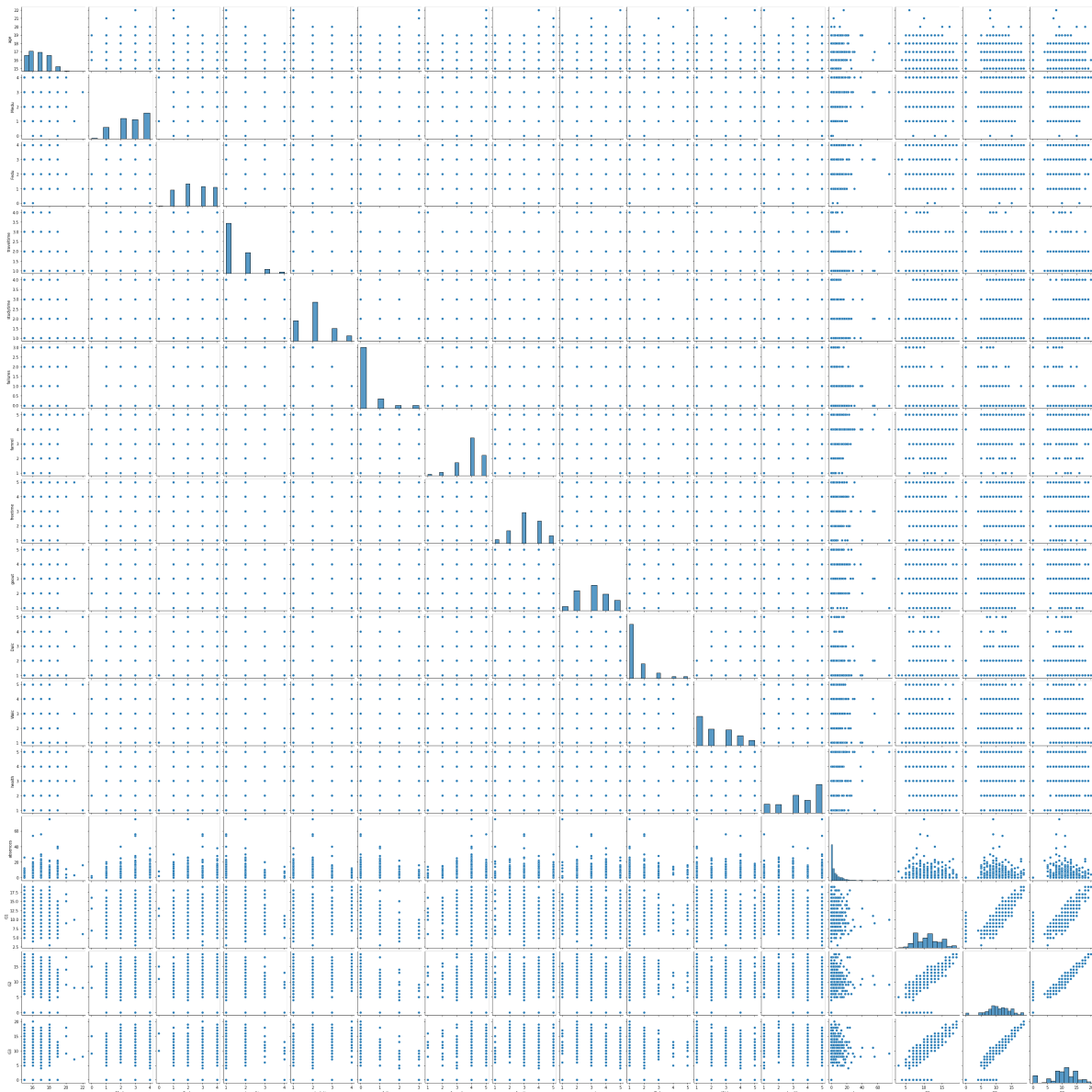

In [25]:

```
plotScatterMatrix(data, 20, 10)
```



In [26]:

```
sns.pairplot(data);
```



In [27]:



```
data.nunique()
```

Out[27]:

```
school      2
sex         2
age         8
address     2
famsize     2
Pstatus     2
Medu       5
Fedu       5
Mjob       5
Fjob       5
reason     4
guardian   3
traveltime 4
studytime  4
failures   4
schoolsup  2
famsup     2
paid       2
activities 2
nursery    2
higher     2
internet   2
romantic   2
famrel     5
freetime   5
goout      5
Dalc       5
Walc       5
health     5
absences   34
G1         17
G2         17
G3         18
dtype: int64
```

In [32]:



```
data['sex'].value_counts()
```

Out[32]:

```
F      208
M      187
Name: sex, dtype: int64
```

In [28]:

```
data['activities'].value_counts()
```

Out[28]:

```
yes    201
no     194
Name: activities, dtype: int64
```

In [29]:

```
data['internet'].value_counts()
```

Out[29]:

```
yes    329
no      66
Name: internet, dtype: int64
```

In [31]:

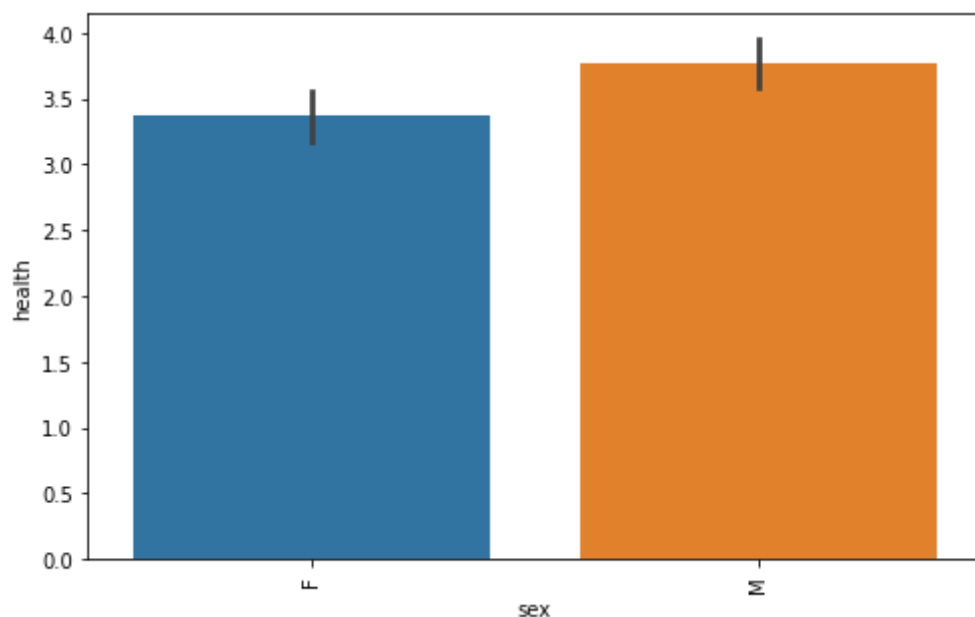
```
data['romantic'].value_counts()
```

Out[31]:

```
no     263
yes    132
Name: romantic, dtype: int64
```

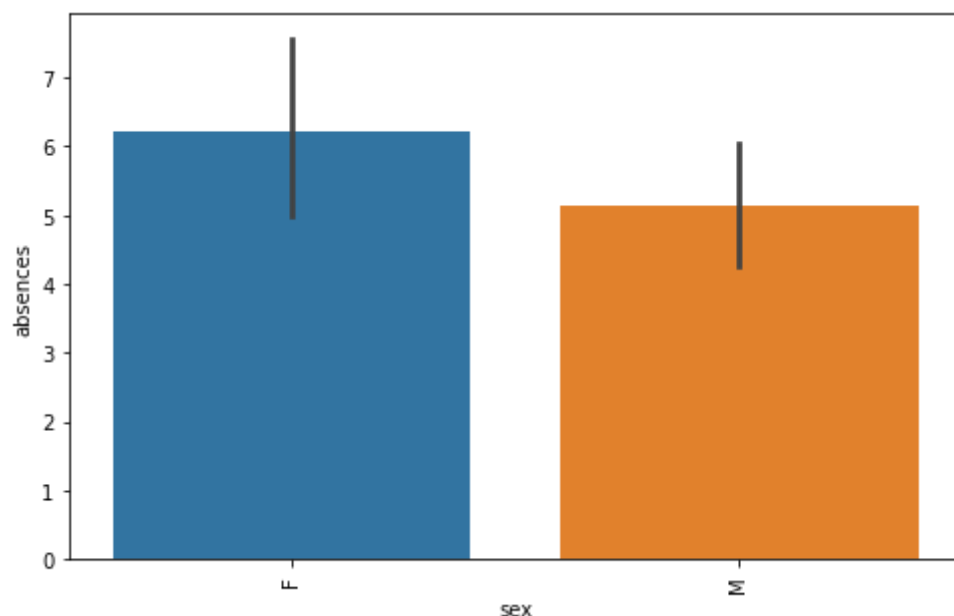
In [34]:

```
plt.figure(figsize=(8,5))
sns.barplot(x = data.sex, y = data.health, data = data)
plt.xticks(rotation = 90)
plt.show()
```



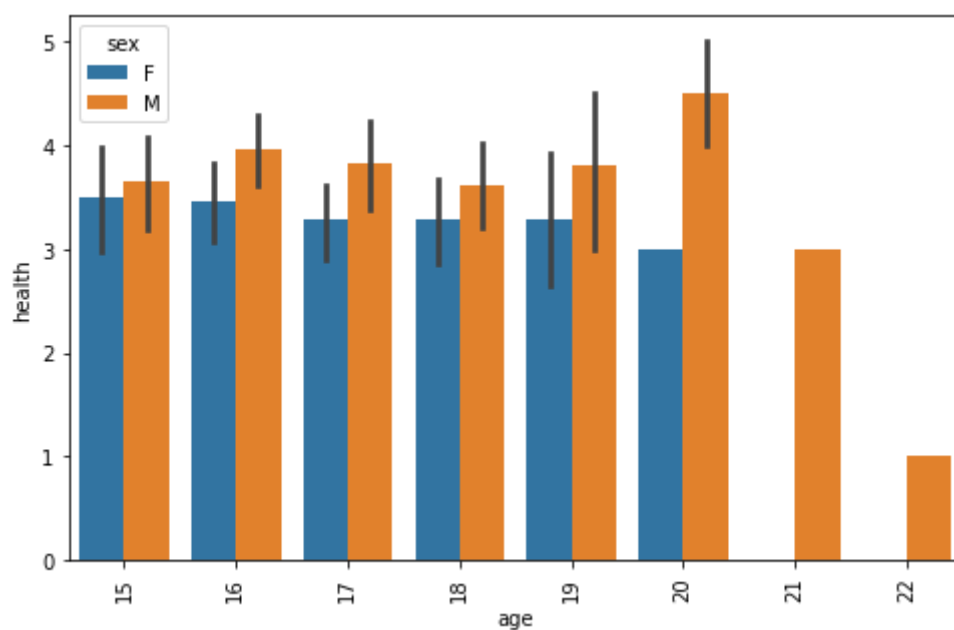
In [35]:

```
plt.figure(figsize=(8,5))
sns.barplot(x = data.sex, y = data.absences, data = data)
plt.xticks(rotation = 90)
plt.show()
```



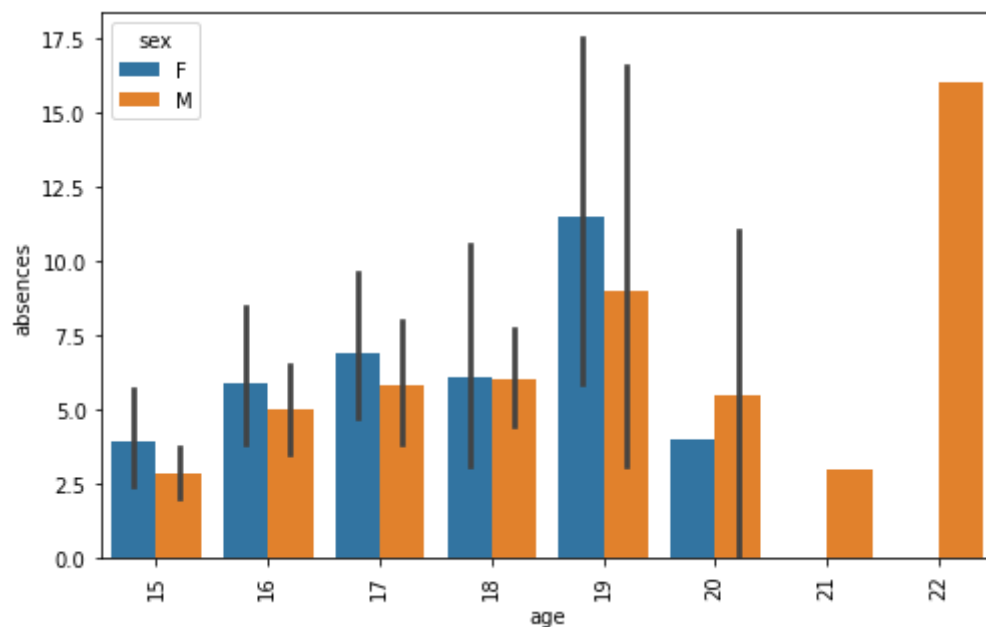
In [36]:

```
plt.figure(figsize=(8,5))
sns.barplot(x = data.age, y = data.health, data = data, hue = data.sex)
plt.xticks(rotation = 90)
plt.show()
```



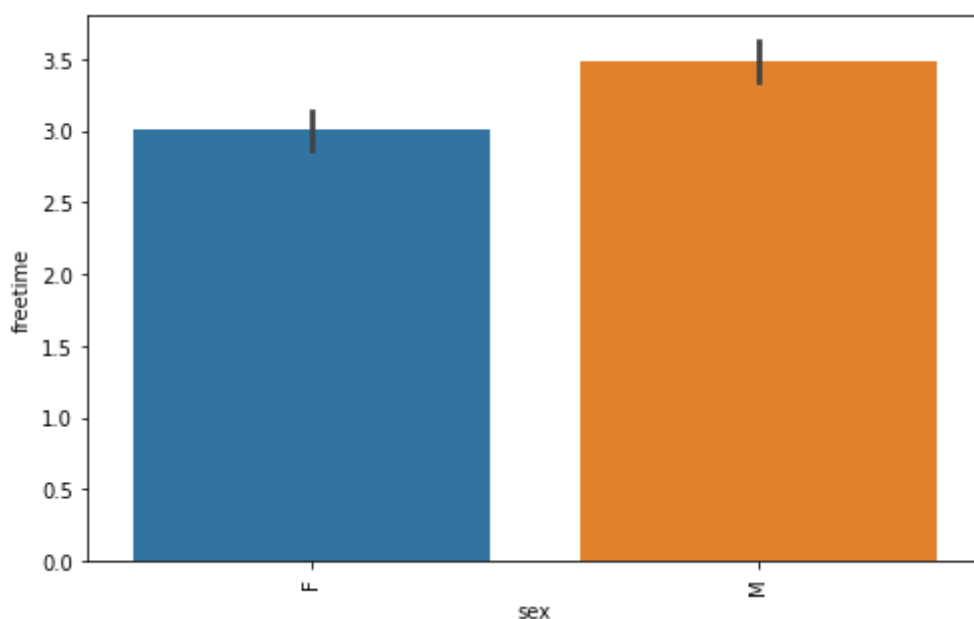
In [37]:

```
plt.figure(figsize=(8,5))
sns.barplot(x = data.age, y = data.absences, data = data, hue = data.sex)
plt.xticks(rotation = 90)
plt.show()
```



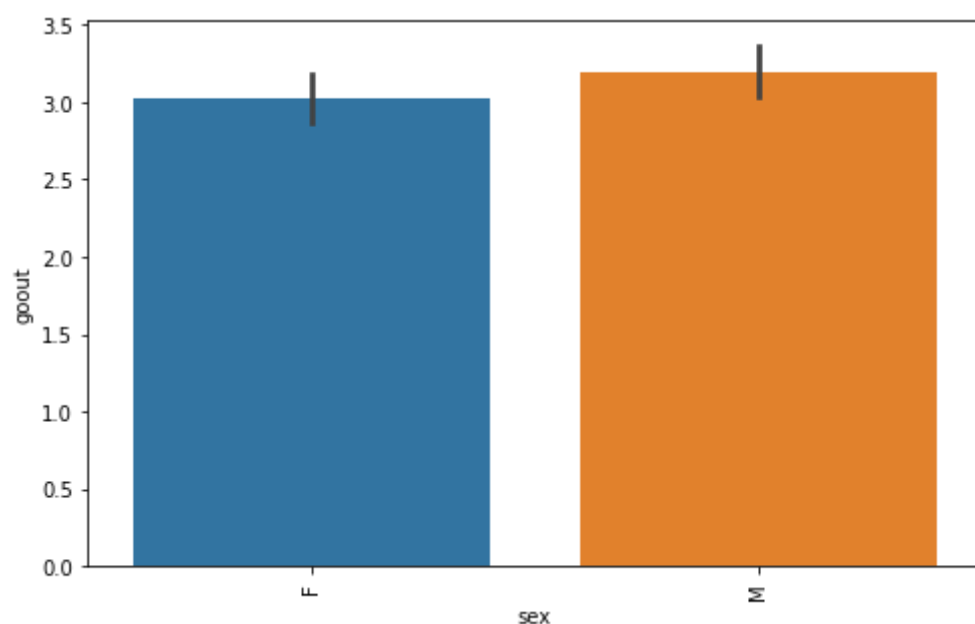
In [38]:

```
plt.figure(figsize=(8,5))
sns.barplot(x = data.sex, y = data.freetime, data = data)
plt.xticks(rotation = 90)
plt.show()
```



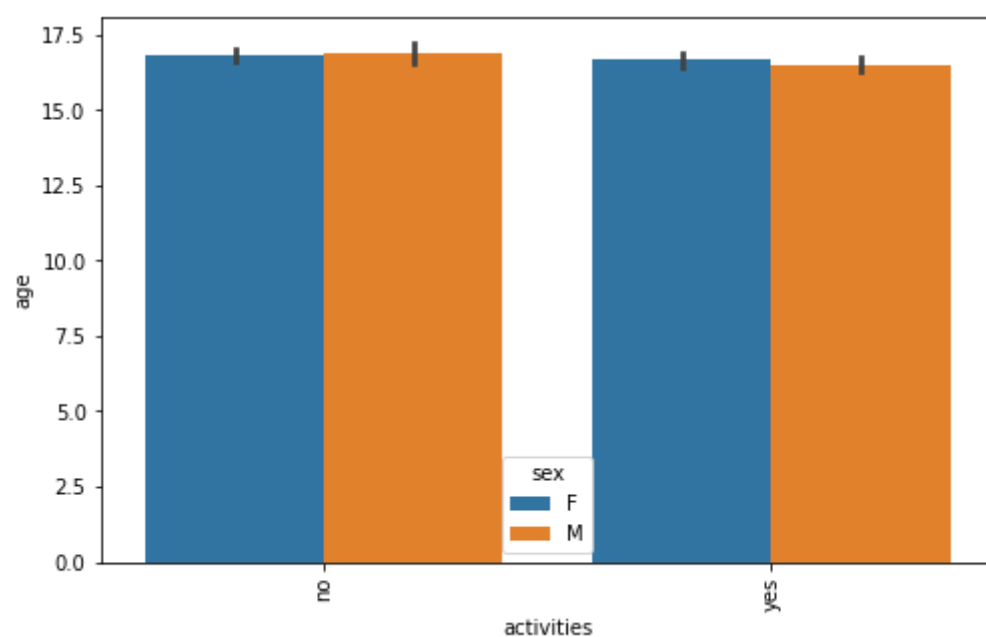
In [39]:

```
plt.figure(figsize=(8,5))  
sns.barplot(x = data.sex, y = data.goout, data = data)  
plt.xticks(rotation = 90)  
plt.show()
```



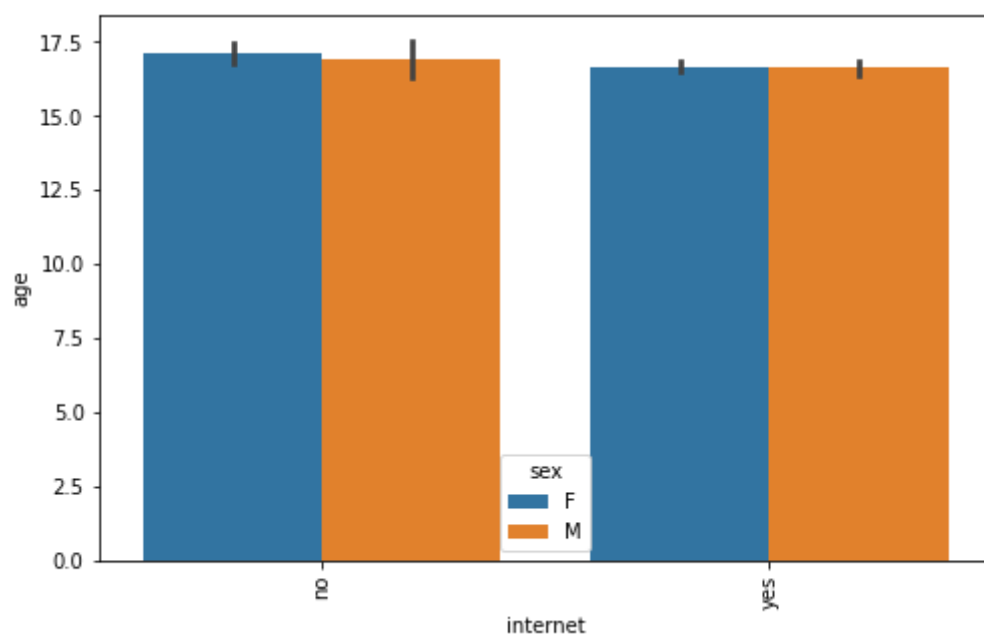
In [41]:

```
plt.figure(figsize=(8,5))  
sns.barplot(x = data.activities, y = data.age, data = data, hue = data.sex)  
plt.xticks(rotation = 90)  
plt.show()
```



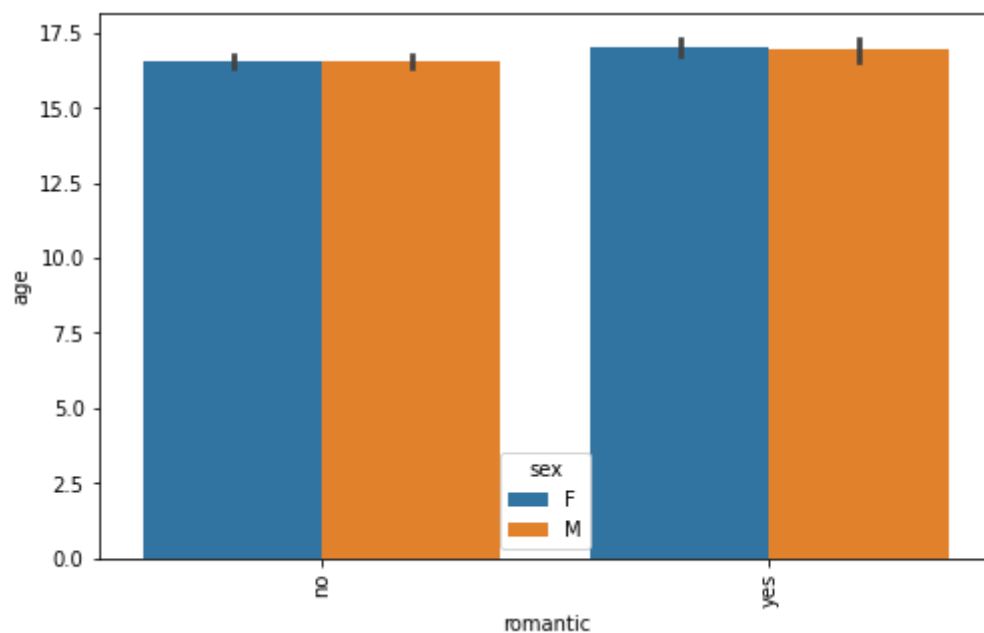
In [42]:

```
plt.figure(figsize=(8,5))  
sns.barplot(x = data.internet, y = data.age, data = data, hue = data.sex)  
plt.xticks(rotation = 90)  
plt.show()
```



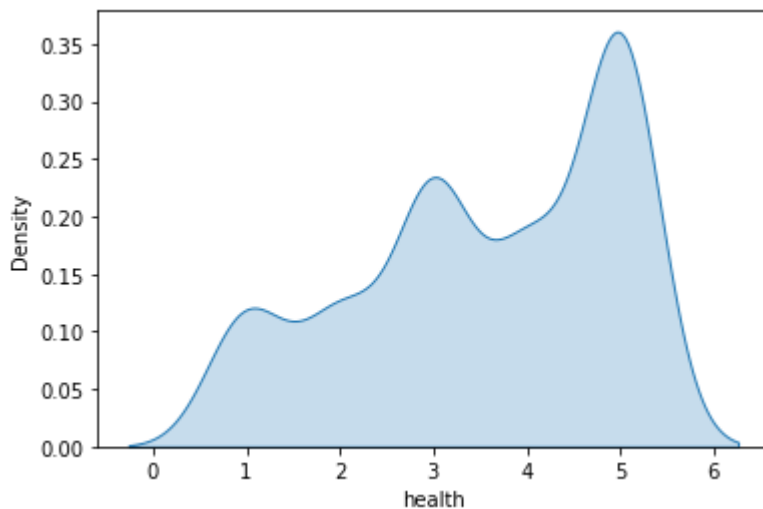
In [43]:

```
plt.figure(figsize=(8,5))  
sns.barplot(x = data.romantic, y = data.age, data = data, hue = data.sex)  
plt.xticks(rotation = 90)  
plt.show()
```



In [44]:

```
sns.kdeplot(x='health', data=data, shade=True);
```



In [45]:

```
data.columns
```

Out[45]:

```
Index(['school', 'sex', 'age', 'address', 'famsize', 'Pstatus', 'Medu', 'Fedu',  
      'Mjob', 'Fjob', 'reason', 'guardian', 'traveltime', 'studytime',  
      'failures', 'schoolsup', 'famsup', 'paid', 'activities', 'nursery',  
      'higher', 'internet', 'romantic', 'famrel', 'freetime', 'goout', 'Dalc',  
      'Walc', 'health', 'absences', 'G1', 'G2', 'G3'],  
      dtype='object')
```

In [46]:

```
new_data=data.drop(columns=['school', 'sex', 'address', 'famsize', 'Pstatus', 'Medu', 'Fedu',  
                           'Mjob', 'Fjob', 'reason', 'guardian', 'schoolsup', 'famsup', 'paid', 'activities',  
                           'higher', 'internet', 'romantic'])
```

In [47]:

```
new_data.head()
```

Out[47]:

	age	travelttime	studytime	failures	famrel	freetime	goout	Dalc	Walc	health	absences	G
0	18	2	2	0	4	3	4	1	1	3	6	
1	17	1	2	0	5	3	3	1	1	3	4	
2	15	1	2	3	4	3	2	2	3	3	10	
3	15	1	3	0	3	2	2	1	1	5	2	1
4	16	1	2	0	4	3	2	1	2	5	4	

In [72]:

```
x = new_data.drop(['G1', 'G2', 'G3'], axis = 1)
y = new_data.G1
```

In [73]:

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
```

In [74]:

```
X_train, X_test, y_train, y_test = train_test_split(x, y, test_size = 0.3)
```

In [75]:

```
model = LinearRegression()
model.fit(X_train, y_train)
```

Out[75]:

```
LinearRegression()
```

In [76]:

```
y_pred = model.predict(X_test)
```

In [77]:

```
print("Training Accuracy :", model.score(X_train, y_train))
print("Testing Accuracy :", model.score(X_test, y_test))
```

```
Training Accuracy : 0.17038753061066803
Testing Accuracy : 0.03497866263159799
```

In [78]:

```
x = new_data.drop(['G1', 'G2', 'G3'], axis = 1)
y = new_data.G2
```

In [79]:

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
```

In [80]:

```
X_train, X_test, y_train, y_test = train_test_split(x, y, test_size = 0.3)
```

In [81]:

```
model1 = LinearRegression()
model1.fit(X_train, y_train)
```

Out[81]:

```
LinearRegression()
```

In [82]:

```
y_pred = model1.predict(X_test)
```

In [83]:

```
print("Training Accuracy :", model1.score(X_train, y_train))
print("Testing Accuracy :", model1.score(X_test, y_test))
```

```
Training Accuracy : 0.17640743035507134
```

```
Testing Accuracy : 0.1321195349551918
```

In [84]:

```
x = new_data.drop(['G1', 'G2', 'G3'], axis = 1)
y = new_data.G3
```

In [85]:

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
```

In [86]:

```
X_train, X_test, y_train, y_test = train_test_split(x, y, test_size = 0.3)
```

In [87]:



```
model2 = LinearRegression()  
model2.fit(X_train, y_train)
```

Out[87]:

```
LinearRegression()
```

In [88]:



```
y_pred = model2.predict(X_test)
```

In [89]:



```
print("Training Accuracy :", model2.score(X_train, y_train))  
print("Testing Accuracy :", model2.score(X_test, y_test))
```

```
Training Accuracy : 0.2062699218905163
```

```
Testing Accuracy : 0.03259803470131295
```