

What Attributes Most Impact Sleep Quality? 😴



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Background

In this project, I will be exploring and analyzing data collected by SleepInc's sleep-tracking app in order to identify lifestyle, health, and demographic factors that most strongly correlate with poor sleep quality.

This will involve the following steps:

- 1. View the data
- 2. Explore the data
- 3. Summarize the data
- 4. Visualize the data
- 5. Make models
- 6. Make and share conclusions

Based on the results of my analysis, I will make conclusions about which factors most strongly correlate with poor sleep quality and make recommendations for people who would like to get a better night's sleep.

The data

SleepInc has provided you with an anonymized dataset of sleep and lifestyle metrics for 374 individuals. This dataset contains average values for each person calculated over the past six months.

The dataset includes 13 columns covering sleep duration, quality, disorders, exercise, stress, diet, demographics, and other factors related to sleep health.

Description	Column
An identifier for each individual.	Person ID
The gender of the person (Male/Female).	Gender
The age of the person in years.	Age
The occupation or profession of the person.	Occupation
The average number of hours the person sleeps per day.	Sleep Duration (hours)
A subjective rating of the quality of sleep, ranging from 1 to 10.	Quality of Sleep (scale: 1-10)
The average number of minutes the person engages in physical activity daily.	Physical Activity Level (minutes/day)
A subjective rating of the stress level experienced by the person, ranging from 1 to 10.	Stress Level (scale: 1-10)
The BMI category of the person (e.g., Underweight, Normal, Overweight).	BMI Category
The average blood pressure measurement of the person, indicated as systolic pressure over diastolic pressure.	Blood Pressure (systolic/diastolic)
The average resting heart rate of the person in beats per minute.	Heart Rate (bpm)
The average number of steps the person takes per day.	Daily Steps
The presence or absence of a sleep disorder in the person (None, Insomnia, Sleep Apnea).	Sleep Disorder

Acknowledgments: Laksika Tharmalingam, Kaggle: https://www.kaggle.com/datasets/uom190346a/sleephealth-and-lifestyle-dataset (https://www.kaggle.com/datasets/uom190346a/sleep-health-and-lifestyle-<u>dataset</u>) (this is a fictitious dataset)

In [52]: # Install common packages import numpy as np import pandas as pd import matplotlib.pyplot as plt

Import and view the data

In [53]: raw_data = pd.read_csv('sleep_health_data.csv')
 raw_data

Out[53]:

	Person ID	Gender	Age	Occupation	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	BMI Category	Blood Pressure	Heart Rate
0	1	Male	27	Software Engineer	6.1	6	42	6	Overweight	126/83	77
1	2	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75
2	3	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75
3	4	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90	85
4	5	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90	85
369	370	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68
370	371	Female	59	Nurse	8.0	9	75	3	Overweight	140/95	68
371	372	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68
372	373	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68
373	374	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68
374 r	374 rows × 13 columns										•

Explore the data

Before performing analyses, I want to get familiar with the dataset. This includes checking for missing values, looking for outliers, and checking out basic statistics.

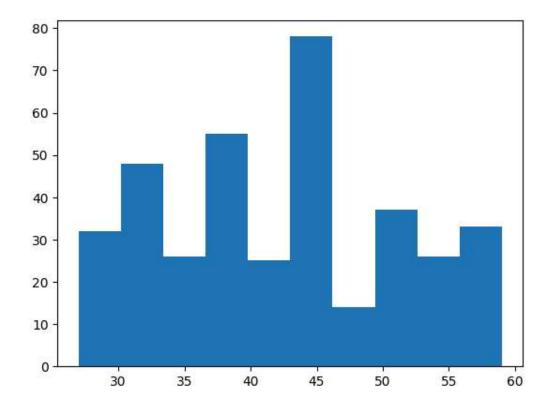
```
# Check for missing values
In [54]:
         raw data.isna().sum()
         # output reveals no missing data
Out[54]: Person ID
                                     0
         Gender
                                     0
         Age
                                     0
         Occupation
                                     0
         Sleep Duration
                                     0
         Quality of Sleep
                                     0
         Physical Activity Level
         Stress Level
                                     0
         BMI Category
                                     0
         Blood Pressure
                                     0
         Heart Rate
                                     0
         Daily Steps
                                     0
         Sleep Disorder
         dtype: int64
         Output reveals no missing data
In [55]: # View histograms/counts for relevant variables
         raw_data['Gender'].value_counts()
Out[55]: Male
                    189
         Female
                    185
         Name: Gender, dtype: int64
         Fairly even split between men and women
         raw_data['Occupation'].value_counts()
In [56]:
Out[56]: Nurse
                                  73
         Doctor
                                  71
         Engineer
                                  63
         Lawyer
                                  47
         Teacher
                                  40
                                  37
         Accountant
         Salesperson
                                  32
         Software Engineer
                                   4
         Scientist
                                   4
         Sales Representative
                                   2
         Manager
         Name: Occupation, dtype: int64
```

Healthcare workers (nurses, doctors) make up the majority of survey participants. Note that this can introduce bias into the analysis as occupations are not evenly represented. Also, since there are only 2 'Sales Representatives', I am going to combine them into the 'Salesperson' category.

```
In [57]:
         raw_data['Occupation'] = raw_data['Occupation'].replace({'Sales Representative': 'Sales
         raw data['Occupation'].value counts()
Out[57]: Nurse
                               73
         Doctor
                               71
                               63
         Engineer
         Lawyer
                               47
         Teacher
                               40
         Accountant
                               37
         Salesperson
                               34
                                4
         Software Engineer
         Scientist
                                4
         Manager
                                1
         Name: Occupation, dtype: int64
```

```
In [58]: plt.hist(raw_data.Age)
```

Out[58]: (array([32., 48., 26., 55., 25., 78., 14., 37., 26., 33.]), array([27., 30.2, 33.4, 36.6, 39.8, 43., 46.2, 49.4, 52.6, 55.8, 59.]), <BarContainer object of 10 artists>)



The majority of participants are in their mid-40's

```
plt.hist(raw_data['Daily Steps'])
In [59]:
Out[59]: (array([ 8.,
                                     8., 69., 69.,
                                                        4., 101.,
                         9., 70.,
                                                                    0., 36.]),
          array([ 3000., 3700., 4400., 5100., 5800., 6500., 7200., 7900.,
                  8600., 9300., 10000.]),
          <BarContainer object of 10 artists>)
           100
            80
            60
            40
            20
                3000
                         4000
                                 5000
                                          6000
                                                  7000
                                                           8000
                                                                   9000
                                                                           10000
         raw data['Sleep Disorder'].value counts(sort=True)
In [60]:
Out[60]: None
                         219
                          78
         Sleep Apnea
                          77
         Insomnia
         Name: Sleep Disorder, dtype: int64
         raw_data['BMI Category'].value_counts()
In [61]:
Out[61]:
         Normal
                          195
         Overweight
                          148
         Normal Weight
                           21
         0bese
                           10
         Name: BMI Category, dtype: int64
In [62]:
         # Combine 'Normal' and 'Normal Weight' categories since they appear to mean the same th
         raw_data['BMI Category'] = raw_data['BMI Category'].replace({'Normal Weight': 'Normal'})
         raw data['BMI Category'].value counts()
Out[62]: Normal
                       216
         Overweight
                       148
         0bese
                         10
         Name: BMI Category, dtype: int64
```

Based on initial exploration, the dataset appears clean and free of any significant outliers. We will next look

Summarize the Data

In [63]: # creating a dataframe of just the numerical variables
data_num = raw_data[['Age', 'Sleep Duration', 'Quality of Sleep', 'Physical Activity Le

In [64]: # View basic stats of all variables in data_num
data_num.describe()

Out[64]:

	Age	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	Heart Rate	Daily Steps
count	374.000000	374.000000	374.000000	374.000000	374.000000	374.000000	374.000000
mean	42.184492	7.132086	7.312834	59.171123	5.385027	70.165775	6816.844920
std	8.673133	0.795657	1.196956	20.830804	1.774526	4.135676	1617.915679
min	27.000000	5.800000	4.000000	30.000000	3.000000	65.000000	3000.000000
25%	35.250000	6.400000	6.000000	45.000000	4.000000	68.000000	5600.000000
50%	43.000000	7.200000	7.000000	60.000000	5.000000	70.000000	7000.000000
75%	50.000000	7.800000	8.000000	75.000000	7.000000	72.000000	8000.000000
max	59.000000	8.500000	9.000000	90.000000	8.000000	86.000000	10000.000000

In [65]: # View mean stats by gender
raw_data.groupby('Gender').mean().sort_values(by='Quality of Sleep')

Out[65]:

	Person ID	Age	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	Heart Rate	Daily Steps
Gende	r							
Mal	e 125.095238	37.074074	7.036508	6.968254	59.201058	6.079365	71.052910	6793.650794
Femal	e 251.254054	47.405405	7.229730	7.664865	59.140541	4.675676	69.259459	6840.540541

The biggest relative differences between genders appears to be quality of sleep and stress level, with females having overall higher quality of sleep and lower stress level.

```
In [66]: # View mean stats by occupation
raw_data.groupby('Occupation').mean().sort_values(by='Quality of Sleep')
```

Out[66]:

	Person ID	Age	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	Heart Rate	Daily Steps
Occupation								
Scientist	75.500000	33.500000	6.000000	5.000000	41.000000	7.000000	78.500000	5350.000000
Salesperson	205.794118	42.617647	6.373529	5.882353	44.117647	7.058824	72.764706	5823.529412
Software Engineer	46.250000	31.250000	6.750000	6.500000	48.000000	6.000000	75.500000	5800.000000
Doctor	64.056338	32.676056	6.970423	6.647887	55.352113	6.732394	71.521127	6808.450704
Teacher	196.075000	41.725000	6.690000	6.975000	45.625000	4.525000	67.225000	5957.500000
Manager	264.000000	45.000000	6.900000	7.000000	55.000000	5.000000	75.000000	5500.000000
Nurse	295.849315	51.794521	7.063014	7.369863	78.589041	5.547945	72.000000	8057.534247
Accountant	153.054054	39.621622	7.113514	7.891892	58.108108	4.594595	68.864865	6881.081081
Lawyer	153.893617	39.425532	7.410638	7.893617	70.425532	5.063830	69.638298	7661,702128
Engineer	245.920635	46.587302	7.987302	8.412698	51.857143	3.888889	67.190476	5980.952381

Scientists have the lowest average quality of sleep, followed by salespersons. At a glance, those who work in sales also appear to have lower sleep durations and higher stress levels.

```
In [67]: # View mean stats by BMI category
raw_data.groupby('BMI Category').mean()
```

Out[67]:

	Person ID	Age	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	Heart Rate	Daily Steps
BN Categor								
Norma	l 139.648148	38.472222	7.387963	7.638889	57.949074	5.134259	68.976852	6875.462963
Obes	e 134.900000	38.000000	6.960000	6.400000	55.000000	5.700000	84.300000	3350.000000
Overweigh	t 260.891892	47.885135	6.770270	6.898649	61.236486	5.729730	70.945946	6965.540541

Participants in the 'Normal' weight category had the highest mean quality of sleep, as well as longest average sleep duration, lower stress levels, and lower heart rate

In [68]: # create a correlation matrix for the numerical variables
 data_num.corr()

Out[68]:

	Age	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	Heart Rate	Daily Steps
Age	1.000000	0.344709	0.473734	0.178993	-0.422344	-0.225606	0.057973
Sleep Duration	0.344709	1.000000	0.883213	0.212360	-0.811023	- 0.516455	-0.039533
Quality of Sleep	0.473734	0.883213	1.000000	0.192896	- 0.898752	-0.659865	0.016791
Physical Activity Level	0.178993	0.212360	0.192896	1.000000	-0.034134	0.136971	0.772723
Stress Level	-0.422344	-0.811023	-0.898752	-0.034134	1.000000	0.670026	0.186829
Heart Rate	-0.225606	-0.516455	-0.659865	0.136971	0.670026	1.000000	-0.030309
Daily Steps	0.057973	-0.039533	0.016791	0.772723	0.186829	-0.030309	1.000000

Based on the correlation matrix, 'quality of sleep' is most correlated with 'sleep duration' and 'stress level'. It appears moderately correlated with 'heart rate'. 'Sleep Duration' and 'Stress Level' are also highly correlated, and 'Stress Level' is moderately correlated with 'Heart Rate'.

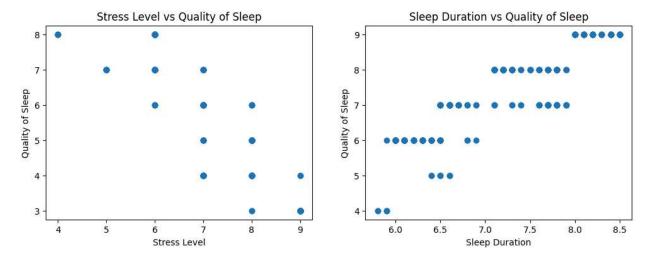
Create Visualizations

```
In [69]: # Visualizing the relationship between the most correlated variables to quality of sleep
import matplotlib.pyplot as plt

fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(12, 4))

# Scatter plot for quality of sleep vs stress level
ax1.scatter(raw_data['Quality of Sleep'], raw_data['Stress Level'])
ax1.set_title('Stress Level vs Quality of Sleep')
ax1.set_ylabel('Stress Level')
ax1.set_ylabel('Quality of Sleep')

# Scatter plot for quality of sleep vs sleep duration
ax2.scatter(raw_data['Sleep Duration'], raw_data['Quality of Sleep'])
ax2.set_title('Sleep Duration vs Quality of Sleep')
ax2.set_ylabel('Sleep Duration')
ax2.set_ylabel('Quality of Sleep')
plt.show()
```



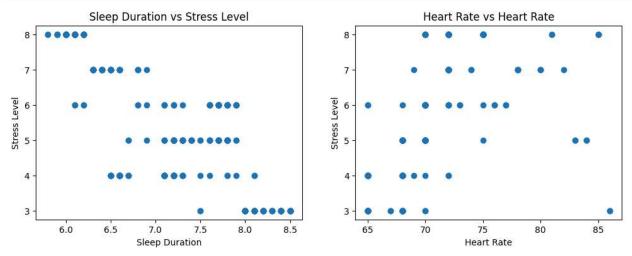
Quality of sleep decreases as stress level goes up, and increases as sleep duration goes up.

```
In [70]: # Visualizing the relationship between stress levels and both sleep duration and heart
fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(12, 4))

# Scatter plot for quality of sleep vs stress level
ax1.scatter(raw_data['Sleep Duration'], raw_data['Stress Level'])
ax1.set_title('Sleep Duration vs Stress Level')
ax1.set_xlabel('Sleep Duration')
ax1.set_ylabel('Stress Level')

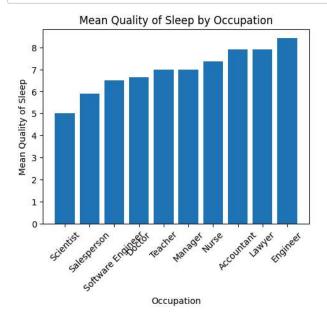
# Scatter plot for quality of sleep vs sleep duration
ax2.scatter(raw_data['Heart Rate'], raw_data['Stress Level'])
ax2.set_title('Heart Rate vs Heart Rate')
ax2.set_xlabel('Heart Rate')
ax2.set_ylabel('Stress Level')

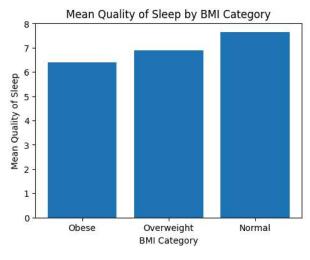
plt.show()
```



Stress level decreases as sleep duration increases, and increases as heart rate also increases.

```
In [71]: # Visualizing some of the categorical variables and their relation to quality of sleep
         fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(12, 4))
         occupation quality = raw data.groupby('Occupation')['Quality of Sleep'].mean().sort val
         bmi quality = raw data.groupby('BMI Category')['Quality of Sleep'].mean().sort values()
         # Bar plot for occupation vs quality of sleep
         ax1.bar(occupation_quality.index, occupation_quality.values)
         ax1.set title('Mean Quality of Sleep by Occupation')
         ax1.set_xlabel('Occupation')
         ax1.set ylabel('Mean Quality of Sleep')
         ax1.set_xticklabels(occupation_quality.index, rotation=45)
         # Bar plot for BMI vs quality of sleep
         ax2.bar(bmi quality.index, bmi quality.values)
         ax2.set title('Mean Quality of Sleep by BMI Category')
         ax2.set_xlabel('BMI Category')
         ax2.set_ylabel('Mean Quality of Sleep')
         plt.show()
```





Create Models

```
In [72]: #Create a linear regression model to predict quality of sleep
from sklearn.linear_model import LinearRegression

# Define the response variable and predictors
X = data_num[['Sleep Duration', 'Stress Level', 'Heart Rate']]
y = data_num['Quality of Sleep']

model = LinearRegression()

# Fit the model
model.fit(X, y)

# Print the coefficients
print('Coefficients:', model.coef_)
```

Coefficients: [0.69333281 -0.29661064 -0.03681626]

```
In [73]: #Create a linear regression model to predict sleep duration
    from sklearn.linear_model import LinearRegression

# Define the response variable and predictors
X = data_num[['Stress Level', 'Heart Rate']]
y = data_num['Sleep Duration']

model = LinearRegression()

# Fit the model
model.fit(X, y)

# Print the coefficients
print('Coefficients:', model.coef_)
```

Coefficients: [-0.37833747 0.00940953]

Conclusion

The linear regression models provide more details about how much our correlated features impact sleep quality. We have the following conclusions:

- 1. The attributes most correlated with sleep quality are sleep duration, stress level, and heart rate. More specifically:
 - For each extra hour of sleep (up to 8.5 hours), quality of sleep improves by about 7%
 - For each 1 point increase in stress level (on a scale of 1-10), quality of sleep decreases by about 3%
 - For each 1 point increase in average heart rate, quality of sleep decreases by about .4%
- 2. Clearly sleep duration is important for improving sleep quality. So what are the factors that impact sleep duration?
 - For each 1 point increase in stress level, sleep duration decreases by about .4 hrs (24 minutes)

- For each 1 point increase in average heart rate sleep duration increases by .009 hrs, or approximately .5 minutes
- 3. In terms of occupation, scientists and salespersons had the lowest quality of sleep ratings, while engineers had the highest.

Recommendations

If you have beens struggling with low quality sleep, consider the following:

- Scientists had an average sleep quality rating of 5 out of 10, compared to engineers who rated their sleep quality at an 8.4 out of 10. That's a pretty significant jump. Is it time to consider a **new career**?
- **Sleep duration** is the most important factor in improving sleep quality. You can increase your sleep duration by **decreasing stress**.
- Though less impactful, heart rate also plays a role in both sleep quality and duration. Finding ways to decrease your average resting heart rate can improve both.