Global Happiness

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Motivation and Summary

- Core Message and Hypothesis
 - There are many factors to take into account when it comes to happiness. On a global scale, how does this measure up between countries?
 - Hypothesis: Countries with a higher happiness rank are more likely to show higher scores in various factors of happiness (GDP per capita, social support, life expectancy, freedom, and generosity), meanwhile creating a larger ecological footprint.
- Questions why did we ask?
 - Do happier countries make more money?
 - Is there a relationship between happiness factors (i.e., social support, life expectancy, freedom, and generosity) and happiness rank?
 - Are "happier" countries more eco-friendly?
- Were we able to answer these questions?
 - The data suggested moderate correlations between some of the variables as shown later on.

Questions and Data

- -The World Happiness Report and the 2016 Global Ecological Footprint were data sources found on kaggle.com.
- -The World Happiness Report provided self-reported happiness scores from their citizens via the Gallup World Poll. Data regarding a country's GDP per capita, their social support, life expectancy, freedom and generosity were assigned an indexed value based on real world data.
- -Data from each year from 2015 through 2019 was provided for close to 160 countries.
- -The 2016 Global Ecological Footprint provided a dataset for 2016 that allowed us to combine it's more in-depth global footprint data with the Happiness Report to further examine if there were correlations with a country's ecological standing.
- -More questions where does the United States rank with the rest of the world? Regions?

Data Cleanup and Exploration

- -Happiness data involved combining and cleaning 5 CSVs
- -Countries data involved renaming countries to match Happiness data
- -Countries data was merged with with Happiness data
- -Obtained top 10 and bottom 10 countries by for each year
- -Compared United States to other countries
- -Grouped Happiness data by region to look for other trends

```
1 # Clean the 2017 data and insert the year
 2 happiness 2017 df = pd.read csv(happiness 2017 file)
 3 happiness 2017 df.insert(0, 'Year', 2017)
 4 happiness 2017 df['Region'] = happiness 2017 df['Country'].map(happiness 2016 df.set index('Country')['Region'])
 5 region = happiness 2017 df['Region']
 6 happiness 2017 df.drop(labels=['Region'], axis=1,inplace = True)
 7 happiness 2017 df.insert(2, 'Region', region)
 8 del happiness 2017 df['Whisker.high']
   del happiness 2017 df['Whisker.low']
   del happiness 2017 df['Dystopia.Residual']
   happiness 2017 df.rename(columns={'Happiness.Rank': 'Happiness Rank',
                                      'Happiness.Score': 'Happiness Score',
12
                                      'Economy..GDP.per.Capita.': 'Economy (GDP per Capita)',
13
                                      'Family': 'Social Support',
14
                                      'Health..Life.Expectancy.': 'Health (Life Expectancy)',
15
                                      'Trust..Government.Corruption.': 'Trust (Government Corruption)'}, inplace=True)
16
   trust = happiness_2017 df['Trust (Government Corruption)']
   happiness 2017 df.drop(labels=['Trust (Government Corruption)'], axis=1,inplace = True)
19 happiness 2017 df.insert(9, 'Trust (Government Corruption)', trust)
20 happiness 2017 df.head()
```

20 mappiness_2017_drinead()											
	Year	Country	Region	Happiness Rank	Happiness Score	Economy (GDP per Capita)	Social Support	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Generosity
0	2017	Norway	Western Europe	1	7.537	1.616463	1.533524	0.796667	0.635423	0.315964	0.362012
1	2017	Denmark	Western Europe	2	7.522	1.482383	1.551122	0.792566	0.626007	0.400770	0.355280
2	2017	Iceland	Western Europe	3	7.504	1.480633	1.610574	0.833552	0.627163	0.153527	0.475540
3	2017	Switzerland	Western Europe	4	7.494	1.564980	1.516912	0.858131	0.620071	0.367007	0.290549
4	2017	Finland	Western Europe	5	7.469	1.443572	1.540247	0.809158	0.617951	0.382612	0.245483

```
happiness df = pd.concat(happiness df list)
   happiness df
   # Export combined dataframe to a csv
   happiness df.to csv('Resources/happiness years.csv', encoding='utf-8', index=False)
1 # Create dataframes for top 10 ranked countries
   top happiness 2015 = happiness df.loc[happiness df['Year'] == 2015, :].head(10)
   top happiness 2015
4 top happiness 2016 = happiness df.loc[happiness df['Year'] == 2016, :].head(10)
5 top happiness 2016
6 top happiness 2017 = happiness df.loc[happiness df['Year'] == 2017, :].head(10)
7 top happiness 2017
8 top happiness 2018 = happiness df.loc[happiness df['Year'] == 2018, :].head(10)
9 top happiness 2018
10 top happiness 2019 = happiness df.loc[happiness df['Year'] == 2019, :].head(10)
11 top happiness 2019
12
   top 10 = [top happiness 2015, top happiness 2016, top happiness 2017, top happiness 2018, top happiness 2019]
```

happiness df list = [happiness 2015 df, happiness 2016 df, happiness 2017 df, happiness 2018 df, happiness 2019 df]

1 # Combine all the year dataframes into a single dataframe

18 top 10 df.to csv('Resources/happiness top 10.csv', encoding='utf-8', index=False)

14

17

16 top 10 df

15 top 10 df = pd.concat(top 10)

3 4 5 6 7	footprint_data1.replace(replacements, regex=True, inplace=True)											
	Country	Region	Population (millions)	HDI	GDP per Capita	A STATE A STATE OF THE STATE OF	Grazing Footprint	Forest Footprint	Carbon Footprint	Fish Footprint		
0	Afghanistan	Middle East/Central Asia	29.82	0.46	\$614.66	0.30	0.20	0.08	0.18	0.00		
1	Albania	Northern/Eastern Europe	3.16	0.73	\$4,534.37	0.78	0.22	0.25	0.87	0.02		
2	Algeria	Africa	38.48	0.73	\$5,430.57	0.60	0.16	0.17	1.14	0.01		
3	Angola	Africa	20.82	0.52	\$4,665.91	0.33	0.15	0.12	0.20	0.09		

0.09 0.78 \$13,205.10

NaN

NaN

NaN

Antigua and

Barbuda

Latin America

Grazing Forest Fishing Urba

Water

0.00

0.07

0.01

0.26

NaN

Lan

0.0

0.0

0.0

0.0

Nai

Land

0.02

0.29

0.03

0.64

NaN

Land

0.20

0.21

0.27

1.42

NaN

Cropland

0.24

0.55

0.24

0.20

NaN

0.00 ...

0.02 ...

0.01 ...

0.09 ...

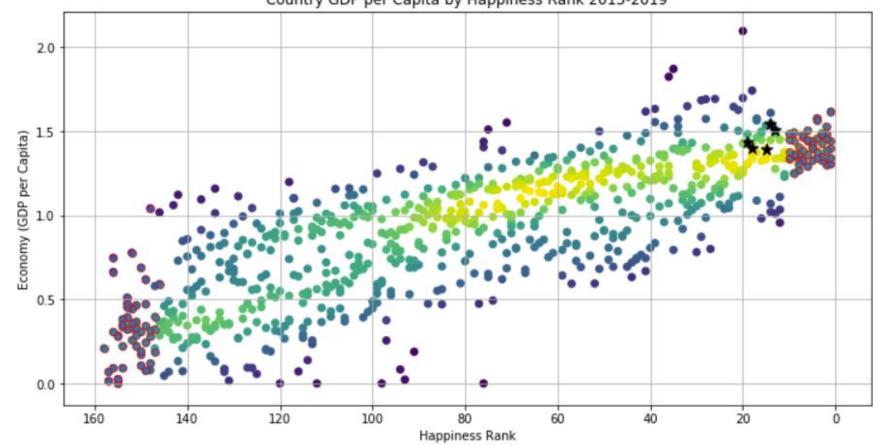
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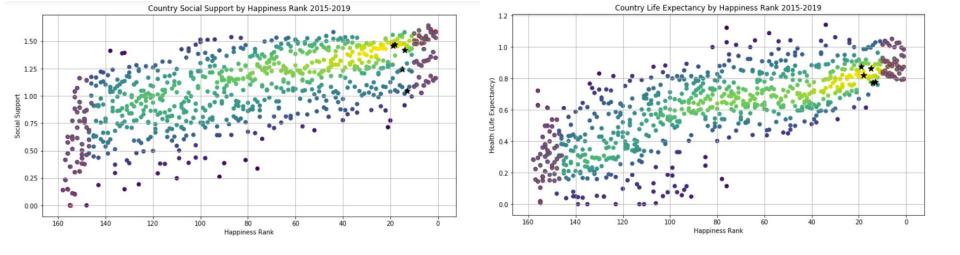
NaN

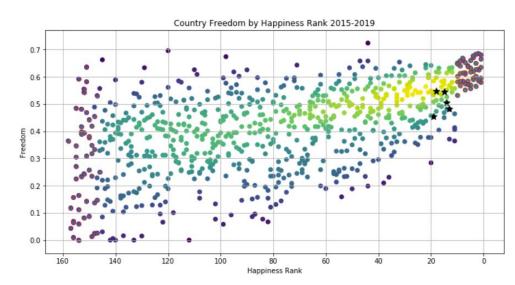
Data Analysis

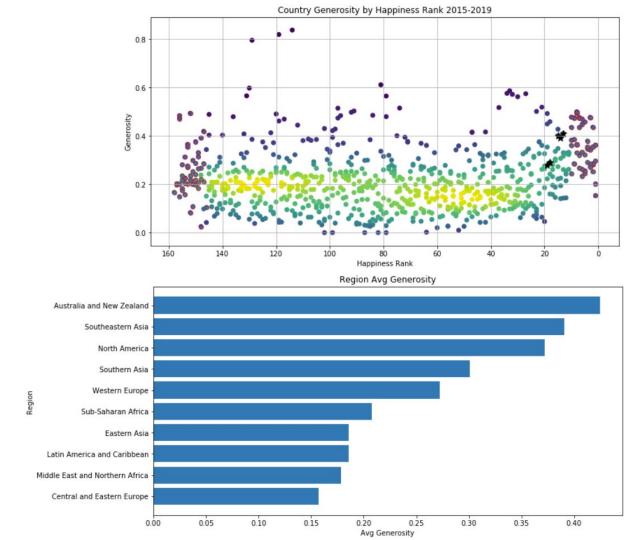
- Steps taken to analyze the data and answer each question you asked in your proposal
 - With the cleaned data, we ran correlation tests
 - We made charts and visuals for the data that showed stronger correlations, or we found surprising
 - United States data points are marked as stars (★) on the charts.

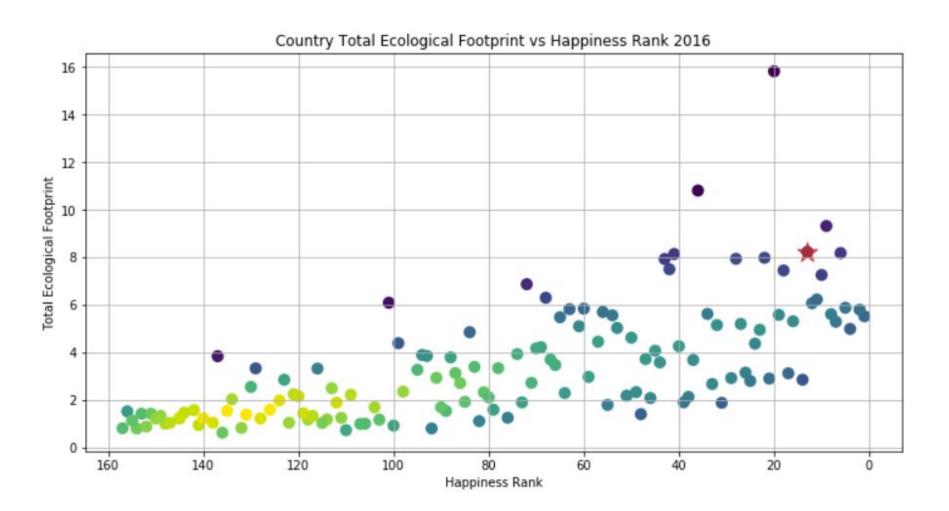
Country GDP per Capita by Happiness Rank 2015-2019

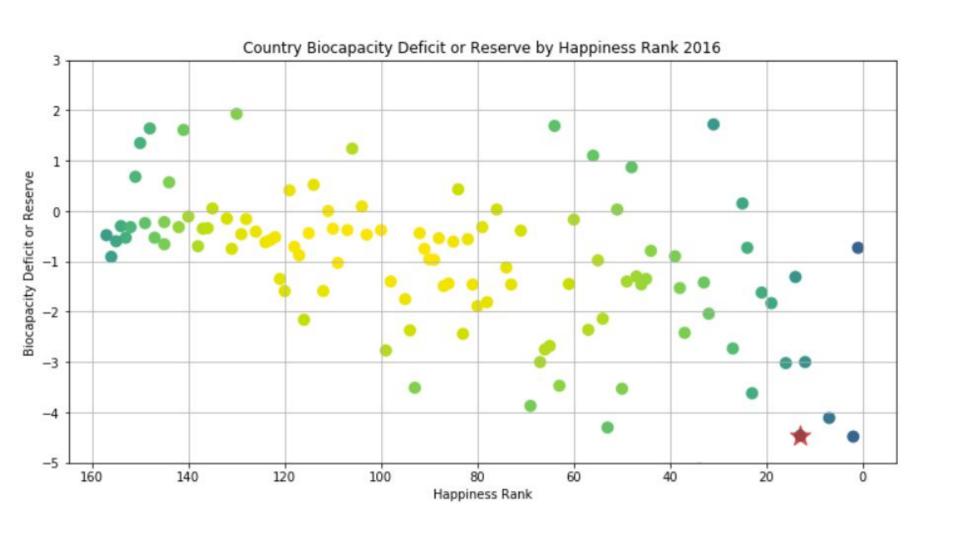












Discussion

- Countries with a higher happiness ranking show a strong positive correlation to having a higher GDP per capita
- The data suggests a moderate, positive correlation between happiness factors (i.e., social support, life expectancy, freedom, and generosity) and happiness rank.
- There is a moderate, positive correlation between happiness rank and total ecological footprint.
 - We were surprised to find that there was not a strong correlation between happiness rank and biocapacity deficit/reserve.

Post Mortem

• Difficulties:

- The data cleanup process took longer than expected
- Focusing our scope on specific questions based on the data we had
- Finding the methods and definitions for the data

Additional Questions:

- Are there correlations between the different factors we analyzed?
- Are the country rankings similar between the different factors?
- Are there different correlations when looking at the outliers?
- How do the factors change for countries over time?

Questions

- -open-floor Q & A with Audience!!!!
- -And GO GIT!!!! WE LOVE IT!!!!