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# **Justin Napolitano Watt Time Take Home**

**Justin Napolitano**

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## Thank You

I am grateful for the opportunity to work on something that I am passionate about. I also had a lot of fun experimenting with the data and learning some new skills along the way.

## About This Document

This document is a sample of the type of work that I like to create. I used the Jupyter Book Packages [Jupyter-Book](#) to build the website and the PDF document. I like working with this suite and Sphinx because it integrates well with Python and Jupyter Notebooks. There are other options available such as Gatsby, but that requires more manual tinkering than I had time to complete.

## Sections

You will find code and no code blog post examples, data exploration/preparation notebooks that I kept to document my thinking, and an analysis of the University of Malaysia paper provided to me.

The scripts I used to prepare the data are found in the data exploration section

Check out my work below.

- Blog Posts
  - *Are We Undercounting Global Methane Emissions? (No Code Example)*
  - *Are We Undercounting Global Methane Emissions? (Code Example)*
- Analysis of The University of Malaysia Paper
  - *Replicating the University of Malaysia Paper.*
  - *Hypothesis Testing the University of Malaysia Paper*
- Data Exploration
  - *Emissions Estimation Data: A Comparison between FAOSTAT and University of Malaysia Estimates*
  - *Merging TRACE and FAOSTAT Data with Geographic Data for Mapping*

## Discussion

### Limiting myself to 300 - 500 words.

I found it difficult to not write a thesis paper. The data exploration notebooks document my journey through the data. My intention was to recreate the Climate Trace article provided to me, test the claims, and review the emissions conversion rates methodology. That work is preserved in the analysis and data exploration sections.

The final post that I prepared feels short, despite not having delved into the depths of the analysis. Nonetheless, I enjoyed the challenge.

## Things I would have Done Differently

I would have begun this process by clearly identifying my audience. Without that in mind, I struggled to formulate an argument.

I would also like to have read the literature to a greater extent in order to cite statements in my post that I felt were under researched.

### Things left to Do

### Data Pipeline

With unlimited time, I would have like to compile my data manipulation processes into a data pipeline program to automate the process without the jupyter overhead. It would better facilitate distributing the repository via github for other researchers to explore.

### Bibliography

Including a bibliography is a feature supported by Jupyter Book. Unfortunately, it can be finicky. I would have liked to include a references section that links to the cited documents.

### 20 and 100 Year Warming Potentials

I would have liked to make an argument supporting including co2equiv in the 20 and 100 year warming potentials measure, but I did not feel authorized to do so. I worried about making an argument that was inaccurate due a lack of exposure to the literature. A full literature review would have been necessary in my opinion to make a valid argument. Unfortunately, I did not have the time nor the space to do so.

### Publish the Website

I chose not to publish the site publicaly at the request of the instructions. I've prepared [methane.jnapolitano.io](https://methane.jnapolitano.io) for deployment to publish this site when permitted to do so.

# **Part I**

## **Blog Posts**





## ARE WE UNDERCOUNTING GLOBAL METHANE EMISSIONS? (NO CODE EXAMPLE)

According to the United Nations Environment Program “methane is the primary contributor to the formation of ground-level ozone.” Over 20 years, “it is 80 times more potent at warming than carbon dioxide.” (UNEP). In fact, one kilogram of CH<sub>4</sub> produced is equivalent to 25 kilograms of CO<sub>2</sub> (Econometra).

The agricultural industry is the primary producer of methane emissions globally. While the cattle industry is a well known culprit, paddy rice cultivation accounts for another 8 per cent of human linked emissions (UNEP).

### 1.1 How are Rice Paddy Emissions Measured?

Methane estimates are calculated by multiplying the hectares of rice paddy in cultivation by a conversion factor.

The Food and Agricultural Organization of the United Nations (FAOSTAT) is the most trusted provider of methane emission estimates. The accuracy of their data however, can be questioned as the organization relies upon official government sources which may be manipulated.

Climate TRACE (Climate TRACE) on the other hand estimates emission levels by calculating the area of cultivated paddies with satellite imaging. While this method may undercount small fields and cultivation at higher altitudes, the metric does not rely upon possibly manipulated sources. This should result in more accurate predictions.

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**Note:** It is important to note that FAOSTAT and TRACE estimate emissions with different conversion factors.

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### 1.2 Differences between FAOSTAT and TRACE Estimates.

#### 1.2.1 Data Review

We will be reviewing the data provided by FAOSTAT and TRACE. The dataframe imported was prepared for this post. You may review the documentation [here](#).

## 1.2.2 Methane Emissions Data 2015 - 2019

	country_name_FAOSTAT	continent	\			
0	Bangladesh	Asia				
1	Brazil	South America				
2	China	Asia				
3	Spain	Europe				
4	Indonesia	Asia				
5	India	Asia				
6	Iran (Islamic Republic of)	Asia				
7	Italy	Europe				
8	Japan	Asia				
9	Cambodia	Asia				
10	Korea (the Republic of)	Asia				
11	Lao People's Democratic Republic (the)	Asia				
12	Sri Lanka	Asia				
13	Myanmar	Asia				
14	Malaysia	Asia				
15	Nepal	Asia				
16	Pakistan	Asia				
17	Philippines (the)	Asia				
18	Korea (the Democratic People's Republic of)	Asia				
19	Thailand	Asia				
20	Taiwan (Province of China)	Asia				
21	United States of America (the)	North America				
22	Viet Nam	Asia				
23	mean	None				
24	total	None				
	CH4_abs_percent_diff_totals	CH4_relative_percent_diff_totals	\			
0	63.425917	-92.881337				
1	93.222402	-174.610414				
2	9.421813	-9.887609				
3	124.705066	76.811284				
4	70.153790	51.936188				
5	32.086209	-38.217479				
6	13.260902	12.436318				
7	80.187551	57.238482				
8	36.874729	31.134372				
9	14.304565	-15.406481				
10	15.813819	14.655057				
11	145.627378	84.268427				
12	27.569003	24.229137				
13	14.697380	-15.863111				
14	13.043224	12.244674				
15	50.437328	40.279401				
16	31.154530	-36.903009				
17	121.748165	75.679166				
18	17.492669	-19.169278				
19	22.551331	20.266184				
20	52.458032	-71.109302				
21	78.850049	56.553728				
22	0.902428	-0.906518				
23	49.129925	3.599038				
24	1129.988279	82.777881				
	CH4_diff_totals	2015	tCH4_2015	2016	tCH4_2016	\

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0	-5.265110e+06	1131293.4	2.344420e+06	1.093480e+06	2.278158e+06
1	-1.096923e+06	138910.3	3.410233e+05	1.262782e+05	3.104189e+05
2	-2.642293e+06	5406593.9	6.133647e+06	5.399920e+06	5.859531e+06
3	2.069382e+05	55082.2	1.141464e+04	5.507310e+04	1.334803e+04
4	6.172158e+06	2407953.5	1.283649e+06	2.387656e+06	1.023129e+06
5	-8.806357e+06	4580248.4	6.219887e+06	4.559136e+06	5.309413e+06
6	6.526191e+04	116486.7	8.774407e+04	1.310085e+05	9.180121e+04
7	3.267972e+05	114574.8	4.995968e+04	1.180030e+05	4.937785e+04
8	5.053990e+05	330353.1	2.305465e+05	3.264030e+05	2.284133e+05
9	-3.570168e+05	436826.0	4.954698e+05	4.590031e+05	5.731698e+05
10	1.169559e+05	167862.2	1.451878e+05	1.635341e+05	1.274597e+05
11	3.747725e+05	94826.8	1.661169e+04	9.563000e+04	1.696441e+04
12	1.337589e+05	132640.0	8.305626e+04	1.217563e+05	1.011743e+05
13	-8.567154e+05	1059409.6	1.132082e+06	1.052288e+06	1.290806e+06
14	7.536891e+04	121942.6	1.057399e+05	1.232328e+05	1.110049e+05
15	3.080032e+05	149262.2	1.007479e+05	1.427237e+05	6.667161e+04
16	-7.340502e+05	383529.3	4.852431e+05	3.813618e+05	5.945922e+05
17	5.944023e+06	1557810.6	3.432021e+05	1.524292e+06	4.073554e+05
18	-8.007815e+04	82823.3	1.143217e+05	8.344230e+04	9.177653e+04
19	1.667678e+06	1554254.0	1.393798e+06	1.703328e+06	1.780993e+06
20	-1.736733e+05	45838.7	7.866956e+04	4.983830e+04	8.089149e+04
21	1.075629e+06	364728.0	1.611324e+05	4.386620e+05	1.618576e+05
22	-6.129998e+04	1381744.4	1.346013e+06	1.365174e+06	1.483777e+06
23	-1.348162e+05	948478.0	9.871116e+05	9.522272e+05	9.587863e+05
24	-3.100772e+06	21814994.0	2.270357e+07	2.190123e+07	2.205208e+07

	2017	tCH4_2017	2017	tCH4_2017	2018 \
0	1.154531e+06	2.098958e+06	1.154531e+06	2.098958e+06	1.144591e+06
1	1.303229e+05	3.725173e+05	1.303229e+05	3.725173e+05	1.216152e+05
2	5.400129e+06	6.355071e+06	5.400129e+06	6.355071e+06	5.302173e+06
3	5.423240e+04	1.217299e+04	5.423240e+04	1.217299e+04	5.292500e+04
4	2.425291e+06	9.615327e+05	2.425291e+06	9.615327e+05	2.405614e+06
5	4.620791e+06	6.228451e+06	4.620791e+06	6.228451e+06	4.661155e+06
6	8.723360e+04	9.620217e+04	8.723360e+04	9.620217e+04	9.393660e+04
7	1.180030e+05	5.443679e+04	1.180030e+05	5.443679e+04	1.094638e+05
8	3.237003e+05	2.708935e+05	3.237003e+05	2.708935e+05	3.222450e+05
9	4.737453e+05	4.517045e+05	4.737453e+05	4.517045e+05	4.793627e+05
10	1.584897e+05	1.463222e+05	1.584897e+05	1.463222e+05	1.549113e+05
11	9.394070e+04	1.168063e+04	9.394070e+04	1.168063e+04	8.333360e+04
12	8.445630e+04	5.911841e+04	8.445630e+04	5.911841e+04	1.110490e+05
13	1.087030e+06	1.205169e+06	1.087030e+06	1.205169e+06	1.118850e+06
14	1.226563e+05	1.111291e+05	1.226563e+05	1.111291e+05	1.252385e+05
15	1.625746e+05	8.081300e+04	1.625746e+05	8.081300e+04	1.538908e+05
16	4.060833e+05	5.372641e+05	4.060833e+05	5.372641e+05	3.934042e+05
17	1.609862e+06	3.836830e+05	1.609862e+06	3.836830e+05	1.606048e+06
18	8.459620e+04	1.085457e+05	8.459620e+04	1.085457e+05	8.394330e+04
19	1.714466e+06	1.164699e+06	1.714466e+06	1.164699e+06	1.702989e+06
20	4.999120e+04	8.705634e+04	4.999120e+04	8.705634e+04	4.941410e+04
21	3.362555e+05	1.684799e+05	3.362555e+05	1.684799e+05	4.121775e+05
22	1.360552e+06	1.406437e+06	1.360552e+06	1.406437e+06	1.336231e+06
23	9.590840e+05	9.727103e+05	9.590840e+05	9.727103e+05	9.575896e+05
24	2.205893e+07	2.237234e+07	2.205893e+07	2.237234e+07	2.202456e+07
	tCH4_2018	2019	tCH4_2019		
0	2.141231e+06	1.144745e+06	2.070985e+06		
1	3.717030e+05	1.110848e+05	3.294713e+05		

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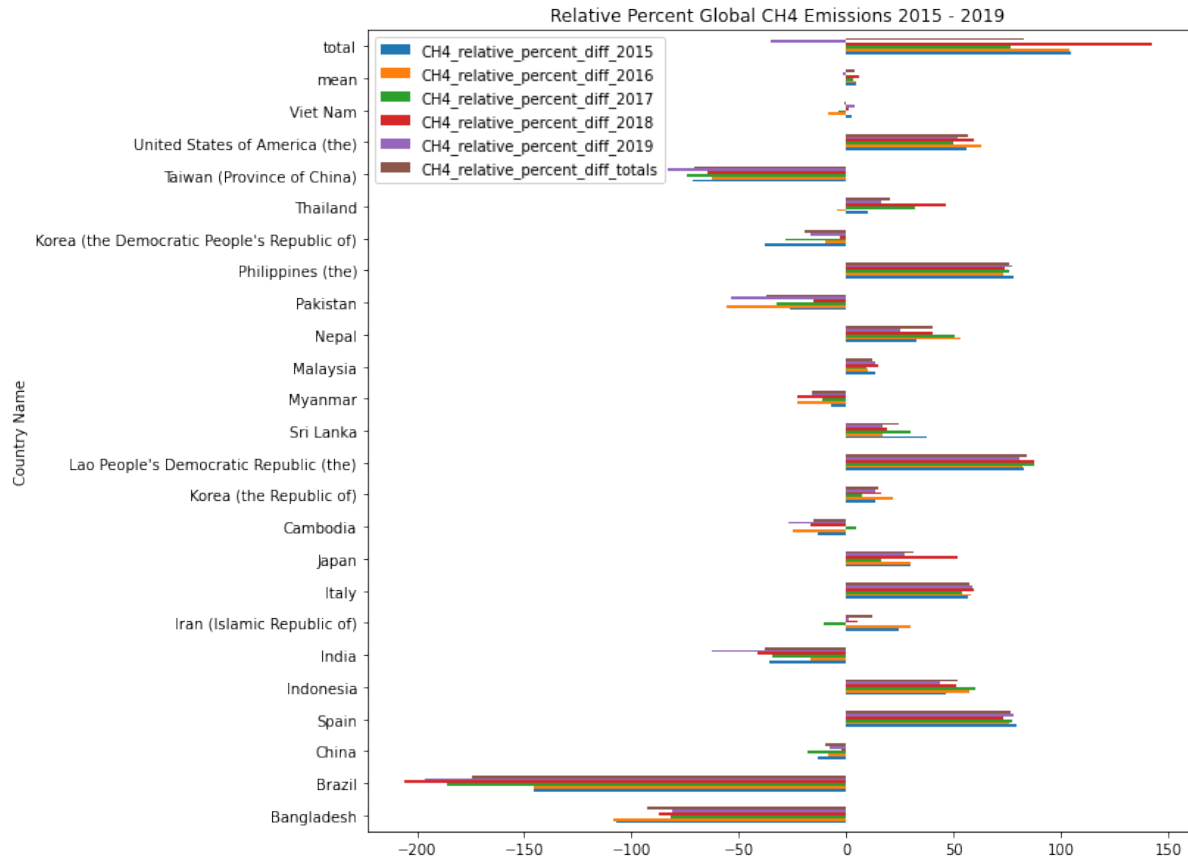
2	5.413962e+06	5.214455e+06	5.603352e+06
3	1.405410e+04	5.209850e+04	1.148324e+04
4	1.176982e+06	2.257604e+06	1.266668e+06
5	6.589798e+06	4.621417e+06	7.501556e+06
6	8.875744e+04	9.610340e+04	9.500199e+04
7	4.469902e+04	1.108951e+05	4.566914e+04
8	1.548252e+05	3.205818e+05	2.332056e+05
9	5.592610e+05	4.683789e+05	5.947277e+05
10	1.293543e+05	1.532609e+05	1.327782e+05
11	1.009675e+04	7.700550e+04	1.461058e+04
12	9.018914e+04	1.021563e+05	8.476088e+04
13	1.372447e+06	1.083100e+06	1.256888e+06
14	1.066525e+05	1.224538e+05	1.056287e+05
15	9.200752e+04	1.562154e+05	1.164235e+05
16	4.532297e+05	4.247551e+05	6.528548e+05
17	4.175210e+05	1.556226e+06	3.584550e+05
18	8.662578e+04	8.293700e+04	9.655062e+04
19	9.166575e+05	1.553836e+06	1.305046e+06
20	8.138151e+04	4.915200e+04	8.990870e+04
21	1.657254e+05	3.501365e+05	1.691351e+05
22	1.317455e+06	1.318431e+06	1.269751e+06
23	9.475920e+05	9.316100e+05	1.017605e+06
24	2.179462e+07	2.142703e+07	2.340491e+07

## Methane Emissions Insights

- FAOSTAT underestimated by about 2.8 percent in comparison to TRACE.
- FAOSTAT underestimated a total of 3,100,772 tonnes of methane over a 5 year period.

Let's review the bar chart below to better understand the data.

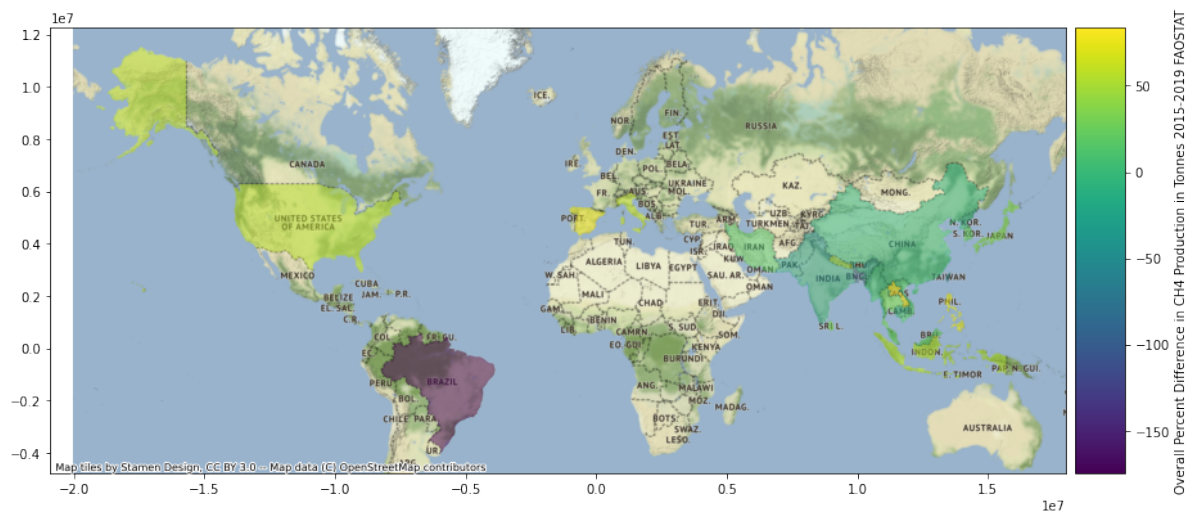
```
<AxesSubplot:title={'center':'Relative Percent Global CH4 Emissions 2015 - 2019'},
  ylabel='Country Name'>
```



### 1.2.3 Major Differences Between FAOSTAT and TRACE for Certain Countries

FaO STAT methods underestimate nearly as often as they overestimate; however, large producer such as Brazil, Bangladesh, China, and India generate large amounts of emissions not considered by FAOSTAT.

The following map illustrates these differences well.



### 1.2.4 FAOSTAT Under Estimation in Brazil China, India, and Bangladesh.

Brazil's emissions are greatly undercounted over the 5 year period. A factor greater than the difference in emission conversion factors would predict. Rice Paddies in cultivation may be hidden from satellites by tree cover or altitude. China, India, and Bangladesh too differ, but at a smaller scale. It is likely that the TRACE data are nearly equivalent to the FAOSTAT data for these contries.

## 1.3 The Case for Climate Trace Estimates

The TRACE and the FAOSTAT methods could both erroneously estimate methane emissions. Climate Trace's advantage is that hecatres of cultivation are estimated from satellite imaging as opposed to government reports. The result of which is a falsifiable report supported by empirical evidence. When calculating the CO2 equivalency of CH4 emissions it is safer to use the TRACE data because the data collection methodology is rigorous.

## 1.4 Estimating CO2 Equivalency

CH4 can be converted to CO2 equivalent by multiplying the 1 kilogram of CH4 by 25 ([Econometra](#)). The dataframe below has been precalculated view the documentation [here](#).

---

**Note:** All differences are recorded as FAOSTAT - TRACE.

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### 1.4.1 Tonnes C02 2015 - 2019 TRACE Estimations

	country_name_TRACE	CO2_abs_percent_diff_totals	\
0	Bangladesh	63.425917	
1	Brazil	93.222402	
2	China	9.421813	
3	Spain	124.705066	
4	Indonesia	70.153790	
5	India	32.086209	
6	Iran (Islamic Republic of)	13.260902	
7	Italy	80.187551	
8	Japan	36.874729	
9	Cambodia	14.304565	
10	Korea (the Republic of)	15.813819	
11	Lao People's Democratic Republic (the)	145.627378	
12	Sri Lanka	27.569003	
13	Myanmar	14.697380	
14	Malaysia	13.043224	
15	Nepal	50.437328	
16	Pakistan	31.154530	
17	Philippines (the)	121.748165	
18	Korea (the Democratic People's Republic of)	17.492669	
19	Thailand	22.551331	
20	Taiwan (Province of China)	52.458032	
21	United States of America (the)	78.850049	
22	Viet Nam	0.902428	
23	mean	49.129925	
24	total	1129.988279	

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	CO2_relative_percent_diff_totals	CO2_diff_totals	tCO2_2015_TRACE	\	
0	-92.881337	-1.316277e+08	5.861049e+07		
1	-174.610414	-2.742306e+07	8.525583e+06		
2	-9.887609	-6.605732e+07	1.533412e+08		
3	76.811284	5.173455e+06	2.853661e+05		
4	51.936188	1.543040e+08	3.209122e+07		
5	-38.217479	-2.201589e+08	1.554972e+08		
6	12.436318	1.631548e+06	2.193602e+06		
7	57.238482	8.169930e+06	1.248992e+06		
8	31.134372	1.263498e+07	5.763662e+06		
9	-15.406481	-8.925421e+06	1.238675e+07		
10	14.655057	2.923897e+06	3.629695e+06		
11	84.268427	9.369313e+06	4.152924e+05		
12	24.229137	3.343972e+06	2.076407e+06		
13	-15.863111	-2.141788e+07	2.830206e+07		
14	12.244674	1.884223e+06	2.643498e+06		
15	40.279401	7.700079e+06	2.518697e+06		
16	-36.903009	-1.835125e+07	1.213108e+07		
17	75.679166	1.486006e+08	8.580052e+06		
18	-19.169278	-2.001954e+06	2.858041e+06		
19	20.266184	4.169196e+07	3.484495e+07		
20	-71.109302	-4.341833e+06	1.966739e+06		
21	56.553728	2.689073e+07	4.028310e+06		
22	-0.906518	-1.532499e+06	3.365033e+07		
23	3.599038	-3.370404e+06	2.467779e+07		
24	82.777881	-7.751930e+07	5.675891e+08		
	tCO2_2016_TRACE	tCO2_2017_TRACE	tCO2_2018_TRACE	tCO2_2019_TRACE	\
0	5.695395e+07	5.247394e+07	5.353076e+07	5.177463e+07	
1	7.760473e+06	9.312934e+06	9.292575e+06	8.236783e+06	
2	1.464883e+08	1.588768e+08	1.353491e+08	1.400838e+08	
3	3.337007e+05	3.043248e+05	3.513524e+05	2.870810e+05	
4	2.557824e+07	2.403832e+07	2.942454e+07	3.166670e+07	
5	1.327353e+08	1.557113e+08	1.647450e+08	1.875389e+08	
6	2.295030e+06	2.405054e+06	2.218936e+06	2.375050e+06	
7	1.234446e+06	1.360920e+06	1.117475e+06	1.141729e+06	
8	5.710333e+06	6.772337e+06	3.870631e+06	5.830141e+06	
9	1.432925e+07	1.129261e+07	1.398153e+07	1.486819e+07	
10	3.186493e+06	3.658056e+06	3.233858e+06	3.319455e+06	
11	4.241102e+05	2.920158e+05	2.524186e+05	3.652645e+05	
12	2.529358e+06	1.477960e+06	2.254728e+06	2.119022e+06	
13	3.227014e+07	3.012923e+07	3.431117e+07	3.142221e+07	
14	2.775123e+06	2.778227e+06	2.666313e+06	2.640717e+06	
15	1.666790e+06	2.020325e+06	2.300188e+06	2.910588e+06	
16	1.486480e+07	1.343160e+07	1.133074e+07	1.632137e+07	
17	1.018389e+07	9.592074e+06	1.043803e+07	8.961374e+06	
18	2.294413e+06	2.713641e+06	2.165645e+06	2.413765e+06	
19	4.452483e+07	2.911748e+07	2.291644e+07	3.262615e+07	
20	2.022287e+06	2.176408e+06	2.034538e+06	2.247717e+06	
21	4.046440e+06	4.211999e+06	4.143136e+06	4.228377e+06	
22	3.709441e+07	3.516092e+07	3.293637e+07	3.174377e+07	
23	2.396966e+07	2.431776e+07	2.368980e+07	2.544012e+07	
24	5.513021e+08	5.593084e+08	5.448654e+08	5.851228e+08	
	Means	Totals			

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0	1.771450e+07	2.733438e+08
1	1.963150e+06	4.312835e+07
2	8.351022e+07	7.341391e+08
3	8.419352e+05	1.561825e+06
4	3.713789e+07	1.427990e+08
5	7.200858e+07	7.962276e+08
6	1.639906e+06	1.148767e+07
7	1.784204e+06	6.103562e+06
8	5.072769e+06	2.794710e+07
9	7.241612e+06	6.685832e+07
10	2.493936e+06	1.702756e+07
11	1.389831e+06	1.749102e+06
12	1.725187e+06	1.045748e+07
13	1.687712e+07	1.564348e+08
14	1.923516e+06	1.350388e+07
15	2.389595e+06	1.141659e+07
16	6.216042e+06	6.807960e+07
17	2.454452e+07	4.775541e+07
18	1.305444e+06	1.244551e+07
19	2.571523e+07	1.640298e+08
20	7.632299e+05	1.044769e+07
21	5.943640e+06	2.065826e+07
22	2.113166e+07	1.705858e+08
23	1.484060e+07	1.220951e+08
24	3.413337e+08	2.808188e+09

---

**Note:** Note a mean of 561,637,600 tonnes CO<sub>2</sub>equiv produced between annually 2015 - 2019.

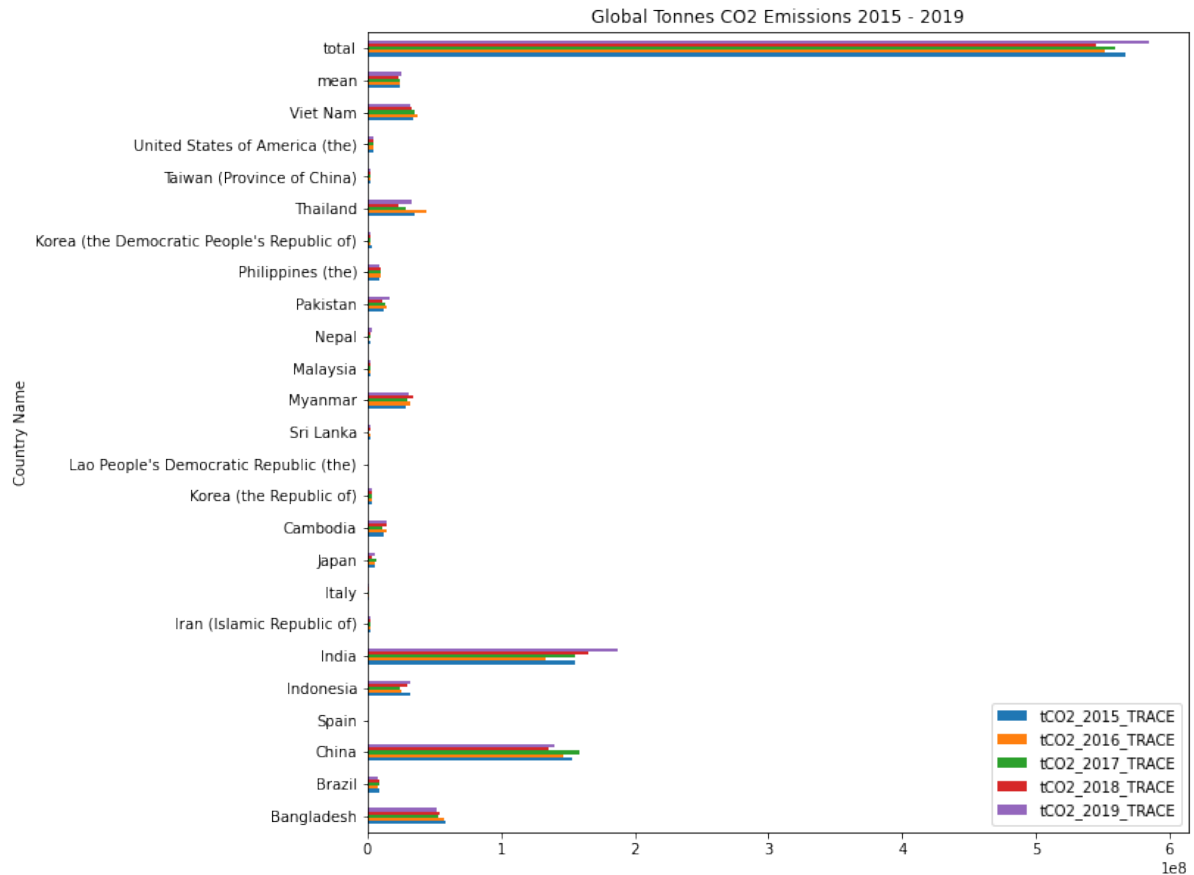
---

### CO<sub>2</sub> Data Insights

- Total difference in 3,370,404 tonnes of CO<sub>2</sub>equiv between the FAOSTAT and TRACE data.
- TRACE reports 2,808,188,000 Tonnes of CO<sub>2</sub>equiv produced between 2015 to 2019.
- There is a relative percent difference of -2.8 percent.

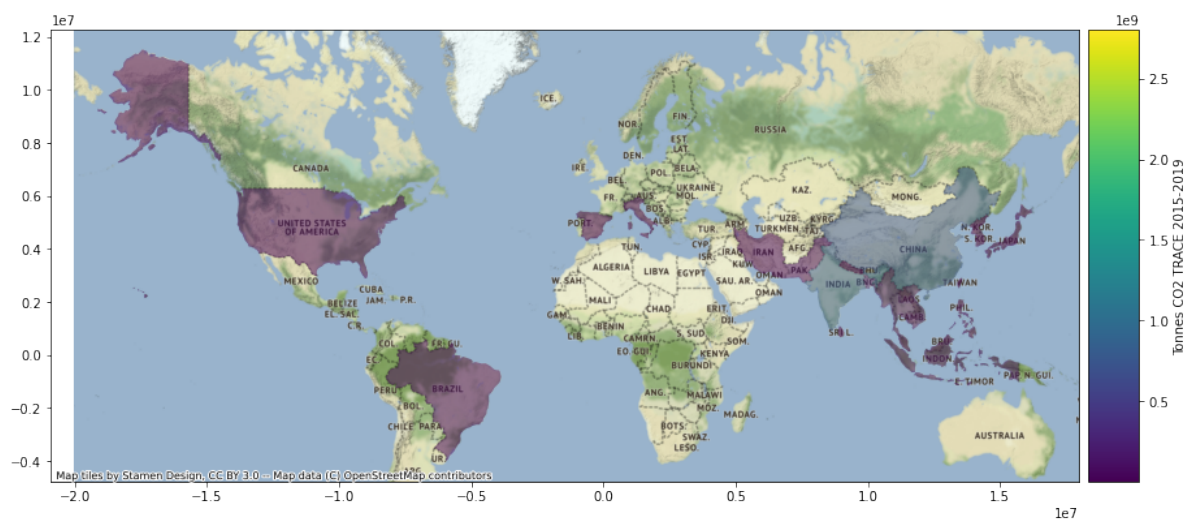
```
<AxesSubplot:title={'center':'Global Tonnes CO2 Emissions 2015 - 2019'}, ylabel=
↪ 'Country Name'>
```





### The Worst Offenders are the Most Dependent on Rice Cultivation

The worst offenders are China, India, and Bangladesh. These states struggle with feeding their massive populations. Including CO2equiv in carbon goals may harm the people of these countries.



---

**Note:** Values in billions.

---

## 1.5 Conclusions

### 1.5.1 FAOSTAT Likely Undercounts Emissions

The TRACE data set on average reports about 2.8 percent more methane emissions than the FAOSTAT data set. This equates to a total methane emissions estimates difference of 3,100,772 tonnes over a 5 year period. The CO<sub>2</sub>equiv over that same period is 77,519,300 tonnes. Neither value is negligible.

### 1.5.2 Countries Should Attempt to Reduce Methane Emissions

The 20 and 100 year warming potentials for methane emission are 84 and 28 [(WG1AR5\_Chapter08, Pg 73)]. These factors will impact our climate if nothing is done. Unfortunately, the countries that produce the most methane from rice cultivation are also those most dependent on the grain. To make the change, realistic solutions must be provided to feed people culturally and geographically dependent on rice cultivation.

## ARE WE UNDERCOUNTING GLOBAL METHANE EMISSIONS? (CODE EXAMPLE)

According to the United Nations Environment Program “methane is the primary contributor to the formation of ground-level ozone.” Over 20 years, “it is 80 times more potent at warming than carbon dioxide.” (UNEP). In fact, one kilogram of CH<sub>4</sub> produced is equivalent to 25 kilograms of CO<sub>2</sub> (Econometra).

The agricultural industry is the primary producer of methane emissions globally. While the cattle industry is a well known culprit, paddy rice cultivation accounts for another 8 per cent of human linked emissions (UNEP).

### 2.1 How are Rice Paddy Emissions Measured?

Methane estimates are calculated by multiplying the hectares of rice paddy in cultivation by a conversion factor.

The Food and Agricultural Organization of the United Nations (FAOSTAT) is the most trusted provider of methane emission estimates. The accuracy of their data however, can be questioned as the organization relies upon official government sources which may be manipulated.

Climate TRACE (Climate TRACE) on the other hand estimates emission levels by calculating the area of cultivated paddies with satellite imaging. While this method may undercount small fields and cultivation at higher altitudes, the metric does not rely upon possibly manipulated sources. This should result in more accurate predictions.

---

**Note:** It is important to note that FAOSTAT and TRACE estimate emissions with different conversion factors.

---

### 2.2 Differences between FAOSTAT and TRACE Estimates.

```
import pandas as pd
import matplotlib.pyplot as plt
import geopandas as gpd
import contextily as cx
import numpy as np
from mpl_toolkits.axes_grid1 import make_axes_locatable
```

## 2.2.1 Data Review

We will be reviewing the data provided by FAOSTAT and TRACE. The dataframe imported was prepared for this post. You may review the documentation [here](#).

```
filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/MERGE_DATA_GEO.geojson"

merge_geo_df = gpd.read_file(filepath)

## Convert to 3857 for easier plotting and spatial math if necessary
merge_geo_df = merge_geo_df.to_crs(epsg=3857)
```

## 2.2.2 Methane Emissions Data 2015 - 2019

```
ch4_data = merge_geo_df[['country_name_FAOSTAT', 'continent', 'CH4_abs_percent_diff_
↳ totals', 'CH4_relative_percent_diff_totals', 'CH4_diff_totals', '2015', 'tCH4_2015',
↳ '2016', 'tCH4_2016', '2017', 'tCH4_2017', '2017', 'tCH4_2017', '2018', 'tCH4_2018',
↳ '2019', 'tCH4_2019']].copy()
ch4_data
```

	country_name_FAOSTAT	continent	\
0	Bangladesh	Asia	
1	Brazil	South America	
2	China	Asia	
3	Spain	Europe	
4	Indonesia	Asia	
5	India	Asia	
6	Iran (Islamic Republic of)	Asia	
7	Italy	Europe	
8	Japan	Asia	
9	Cambodia	Asia	
10	Korea (the Republic of)	Asia	
11	Lao People's Democratic Republic (the)	Asia	
12	Sri Lanka	Asia	
13	Myanmar	Asia	
14	Malaysia	Asia	
15	Nepal	Asia	
16	Pakistan	Asia	
17	Philippines (the)	Asia	
18	Korea (the Democratic People's Republic of)	Asia	
19	Thailand	Asia	
20	Taiwan (Province of China)	Asia	
21	United States of America (the)	North America	
22	Viet Nam	Asia	
23	mean	None	
24	total	None	

	CH4_abs_percent_diff_totals	CH4_relative_percent_diff_totals	\
0	63.425917	-92.881337	
1	93.222402	-174.610414	
2	9.421813	-9.887609	
3	124.705066	76.811284	
4	70.153790	51.936188	
5	32.086209	-38.217479	

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6	13.260902	12.436318
7	80.187551	57.238482
8	36.874729	31.134372
9	14.304565	-15.406481
10	15.813819	14.655057
11	145.627378	84.268427
12	27.569003	24.229137
13	14.697380	-15.863111
14	13.043224	12.244674
15	50.437328	40.279401
16	31.154530	-36.903009
17	121.748165	75.679166
18	17.492669	-19.169278
19	22.551331	20.266184
20	52.458032	-71.109302
21	78.850049	56.553728
22	0.902428	-0.906518
23	49.129925	3.599038
24	1129.988279	82.777881

CH4_diff_totals	2015	tCH4_2015	2016	tCH4_2016	\
0	-5.265110e+06	1131293.4	2.344420e+06	1.093480e+06	2.278158e+06
1	-1.096923e+06	138910.3	3.410233e+05	1.262782e+05	3.104189e+05
2	-2.642293e+06	5406593.9	6.133647e+06	5.399920e+06	5.859531e+06
3	2.069382e+05	55082.2	1.141464e+04	5.507310e+04	1.334803e+04
4	6.172158e+06	2407953.5	1.283649e+06	2.387656e+06	1.023129e+06
5	-8.806357e+06	4580248.4	6.219887e+06	4.559136e+06	5.309413e+06
6	6.526191e+04	116486.7	8.774407e+04	1.310085e+05	9.180121e+04
7	3.267972e+05	114574.8	4.995968e+04	1.180030e+05	4.937785e+04
8	5.053990e+05	330353.1	2.305465e+05	3.264030e+05	2.284133e+05
9	-3.570168e+05	436826.0	4.954698e+05	4.590031e+05	5.731698e+05
10	1.169559e+05	167862.2	1.451878e+05	1.635341e+05	1.274597e+05
11	3.747725e+05	94826.8	1.661169e+04	9.563000e+04	1.696441e+04
12	1.337589e+05	132640.0	8.305626e+04	1.217563e+05	1.011743e+05
13	-8.567154e+05	1059409.6	1.132082e+06	1.052288e+06	1.290806e+06
14	7.536891e+04	121942.6	1.057399e+05	1.232328e+05	1.110049e+05
15	3.080032e+05	149262.2	1.007479e+05	1.427237e+05	6.667161e+04
16	-7.340502e+05	383529.3	4.852431e+05	3.813618e+05	5.945922e+05
17	5.944023e+06	1557810.6	3.432021e+05	1.524292e+06	4.073554e+05
18	-8.007815e+04	82823.3	1.143217e+05	8.344230e+04	9.177653e+04
19	1.667678e+06	1554254.0	1.393798e+06	1.703328e+06	1.780993e+06
20	-1.736733e+05	45838.7	7.866956e+04	4.983830e+04	8.089149e+04
21	1.075629e+06	364728.0	1.611324e+05	4.386620e+05	1.618576e+05
22	-6.129998e+04	1381744.4	1.346013e+06	1.365174e+06	1.483777e+06
23	-1.348162e+05	948478.0	9.871116e+05	9.522272e+05	9.587863e+05
24	-3.100772e+06	21814994.0	2.270357e+07	2.190123e+07	2.205208e+07

2017	tCH4_2017	2017	tCH4_2017	2018	\
0	1.154531e+06	2.098958e+06	1.154531e+06	2.098958e+06	1.144591e+06
1	1.303229e+05	3.725173e+05	1.303229e+05	3.725173e+05	1.216152e+05
2	5.400129e+06	6.355071e+06	5.400129e+06	6.355071e+06	5.302173e+06
3	5.423240e+04	1.217299e+04	5.423240e+04	1.217299e+04	5.292500e+04
4	2.425291e+06	9.615327e+05	2.425291e+06	9.615327e+05	2.405614e+06
5	4.620791e+06	6.228451e+06	4.620791e+06	6.228451e+06	4.661155e+06
6	8.723360e+04	9.620217e+04	8.723360e+04	9.620217e+04	9.393660e+04
7	1.180030e+05	5.443679e+04	1.180030e+05	5.443679e+04	1.094638e+05

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8	3.237003e+05	2.708935e+05	3.237003e+05	2.708935e+05	3.222450e+05
9	4.737453e+05	4.517045e+05	4.737453e+05	4.517045e+05	4.793627e+05
10	1.584897e+05	1.463222e+05	1.584897e+05	1.463222e+05	1.549113e+05
11	9.394070e+04	1.168063e+04	9.394070e+04	1.168063e+04	8.333360e+04
12	8.445630e+04	5.911841e+04	8.445630e+04	5.911841e+04	1.110490e+05
13	1.087030e+06	1.205169e+06	1.087030e+06	1.205169e+06	1.118850e+06
14	1.226563e+05	1.111291e+05	1.226563e+05	1.111291e+05	1.252385e+05
15	1.625746e+05	8.081300e+04	1.625746e+05	8.081300e+04	1.538908e+05
16	4.060833e+05	5.372641e+05	4.060833e+05	5.372641e+05	3.934042e+05
17	1.609862e+06	3.836830e+05	1.609862e+06	3.836830e+05	1.606048e+06
18	8.459620e+04	1.085457e+05	8.459620e+04	1.085457e+05	8.394330e+04
19	1.714466e+06	1.164699e+06	1.714466e+06	1.164699e+06	1.702989e+06
20	4.999120e+04	8.705634e+04	4.999120e+04	8.705634e+04	4.941410e+04
21	3.362555e+05	1.684799e+05	3.362555e+05	1.684799e+05	4.121775e+05
22	1.360552e+06	1.406437e+06	1.360552e+06	1.406437e+06	1.336231e+06
23	9.590840e+05	9.727103e+05	9.590840e+05	9.727103e+05	9.575896e+05
24	2.205893e+07	2.237234e+07	2.205893e+07	2.237234e+07	2.202456e+07

	tCH4_2018	2019	tCH4_2019
0	2.141231e+06	1.144745e+06	2.070985e+06
1	3.717030e+05	1.110848e+05	3.294713e+05
2	5.413962e+06	5.214455e+06	5.603352e+06
3	1.405410e+04	5.209850e+04	1.148324e+04
4	1.176982e+06	2.257604e+06	1.266668e+06
5	6.589798e+06	4.621417e+06	7.501556e+06
6	8.875744e+04	9.610340e+04	9.500199e+04
7	4.469902e+04	1.108951e+05	4.566914e+04
8	1.548252e+05	3.205818e+05	2.332056e+05
9	5.592610e+05	4.683789e+05	5.947277e+05
10	1.293543e+05	1.532609e+05	1.327782e+05
11	1.009675e+04	7.700550e+04	1.461058e+04
12	9.018914e+04	1.021563e+05	8.476088e+04
13	1.372447e+06	1.083100e+06	1.256888e+06
14	1.066525e+05	1.224538e+05	1.056287e+05
15	9.200752e+04	1.562154e+05	1.164235e+05
16	4.532297e+05	4.247551e+05	6.528548e+05
17	4.175210e+05	1.556226e+06	3.584550e+05
18	8.662578e+04	8.293700e+04	9.655062e+04
19	9.166575e+05	1.553836e+06	1.305046e+06
20	8.138151e+04	4.915200e+04	8.990870e+04
21	1.657254e+05	3.501365e+05	1.691351e+05
22	1.317455e+06	1.318431e+06	1.269751e+06
23	9.475920e+05	9.316100e+05	1.017605e+06
24	2.179462e+07	2.142703e+07	2.340491e+07

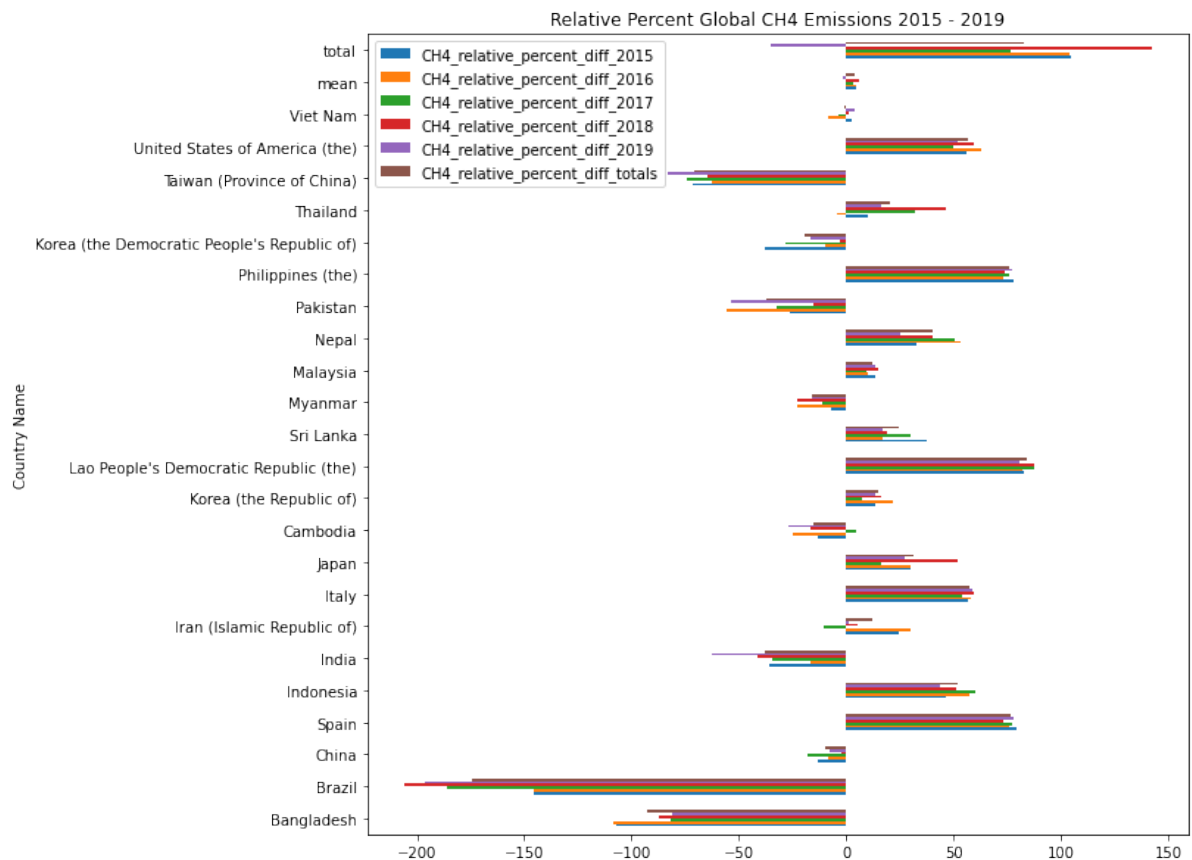
## Methane Emissions Insights

- FAOSTAT underestimated by about 2.8 percent in comparison to TRACE.
- FAOSTAT underestimated a total of 3,100,772 tonnes of methane over a 5 year period.

Let's review the bar chart below to better understand the data.

```
merge_geo_df.plot(kind = "barh", x = 'country_name_FAOSTAT', y = ['CH4_relative_
percent_diff_2015', 'CH4_relative_percent_diff_2016', 'CH4_relative_percent_diff_2017
', 'CH4_relative_percent_diff_2018', 'CH4_relative_percent_diff_2019', 'CH4_relative_
percent_diff_totals'], xlabel = "Country Name", ylabel = "Tonnes CH4", title =
"Relative Percent Global CH4 Emissions 2015 - 2019", figsize = (10,10))
```

```
<AxesSubplot:title={'center':'Relative Percent Global CH4 Emissions 2015 - 2019'},
ylabel='Country Name'>
```

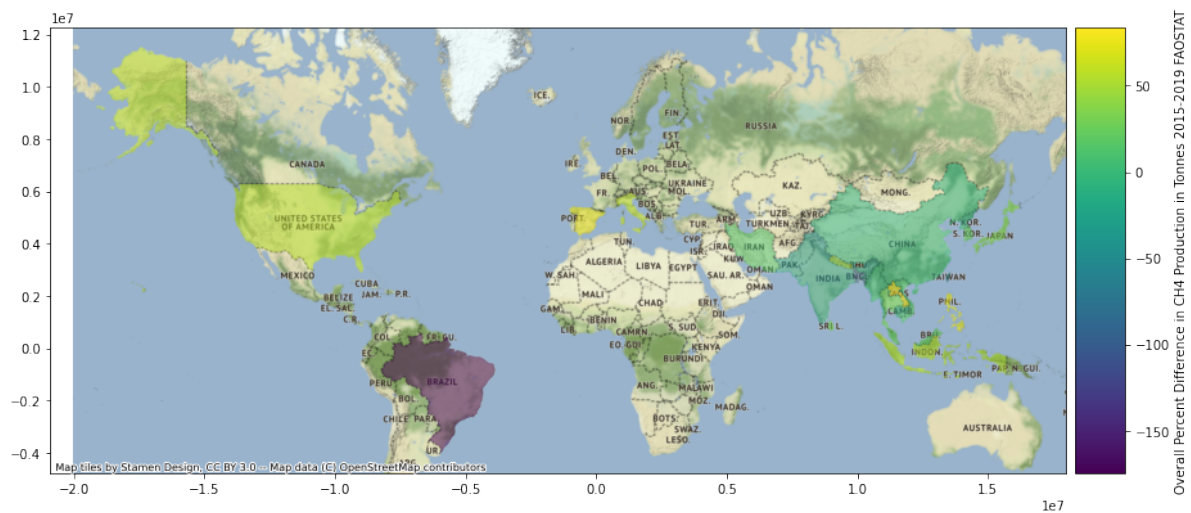


## 2.2.3 Major Differences Between FAOSTAT and TRACE for Certain Countries

FaoSTAT methods underestimate nearly as often as they overestimate; however, large producer such as Brazil, Bangladesh, China, and India generate large amounts of emissions not considered by FAOSTAT.

The following map illustrates these differences well.

```
fig, ax = plt.subplots(1, 1, figsize=(15,15))
divider = make_axes_locatable(ax)
cax = divider.append_axes("right", size="5%", pad=0.1)
merge_map = merge_geo_df.plot(column='CH4_relative_percent_diff_totals', ax=ax,
    legend=True, cax=cax, alpha = .5, legend_kwds={'label': "Overall Percent Difference",
    in CH4 Production in Tonnes 2015-2019 FAOSTAT",
    'orientation': "vertical"})
cx.add_basemap(merge_map, zoom=3)
```



## 2.2.4 FAOSTAT Under Estimation in Brazil China, India, and Bangladesh.

Brazil's emissions are greatly undercounted over the 5 year period. A factor greater than the difference in emission conversion factors would predict. Rice Paddies in cultivation may be hidden from satellites by tree cover or altitude. China, India, and Bangladesh too differ, but at a smaller scale. It is likely that the TRACE data are nearly equivalent to the FAOSTAT data for these countries.

## 2.3 The Case for Climate Trace Estimates

The TRACE and the FAOSTAT methods could both erroneously estimate methane emissions. Climate Trace's advantage is that hectares of cultivation are estimated from satellite imaging as opposed to government reports. The result of which is a falsifiable report supported by empirical evidence. When calculating the CO2 equivalency of CH4 emissions it is safer to use the TRACE data because the data collection methodology is rigorous.



## 2.4 Estimating CO2 Equivalency

CH4 can be converted to CO2 equivalent by multiplying the 1 kilogram of CH4 by 25 (Econometra). The dataframe below has been precalculated view the documentation [here](#).

```
co2_data = merge_geo_df[['country_name_TRACE', 'continent', 'CO2_abs_percent_diff_
↪totals', 'CO2_relative_percent_diff_totals', 'CO2_diff_totals', 'tCO2_2015_FAOSTAT',
↪tCO2_2015_TRACE', 'tCO2_2016_FAOSTAT', 'tCO2_2016_TRACE', 'tCO2_2017_FAOSTAT',
↪tCO2_2017_TRACE', 'tCO2_2018_FAOSTAT', 'tCO2_2018_TRACE', 'tCO2_2019_FAOSTAT', 'tCO2_
↪2019_TRACE']].copy()
```

**Note:** All differences are recorded as FAOSTAT - TRACE.

### 2.4.1 Tonnes CO2 2015 - 2019 TRACE Estimations

```
TRACE_CO2_df = merge_geo_df[['country_name_TRACE', 'CO2_abs_percent_diff_totals', 'CO2_
↪relative_percent_diff_totals', 'CO2_diff_totals', 'tCO2_2015_TRACE', 'tCO2_2016_TRACE
↪', 'tCO2_2017_TRACE', 'tCO2_2018_TRACE', 'tCO2_2019_TRACE']].copy()
TRACE_CO2_df['Means'] = TRACE_CO2_df.select_dtypes(np.number).mean(axis=1)
TRACE_CO2_df['Totals'] = TRACE_CO2_df[['country_name_TRACE', 'tCO2_2015_TRACE', 'tCO2_
↪2016_TRACE', 'tCO2_2017_TRACE', 'tCO2_2018_TRACE', 'tCO2_2019_TRACE']].select_
↪dtypes(np.number).sum(axis=1)
TRACE_CO2_df
```

	country_name_TRACE	CO2_abs_percent_diff_totals	\
0	Bangladesh	63.425917	
1	Brazil	93.222402	
2	China	9.421813	
3	Spain	124.705066	
4	Indonesia	70.153790	
5	India	32.086209	
6	Iran (Islamic Republic of)	13.260902	
7	Italy	80.187551	
8	Japan	36.874729	
9	Cambodia	14.304565	
10	Korea (the Republic of)	15.813819	
11	Lao People's Democratic Republic (the)	145.627378	
12	Sri Lanka	27.569003	
13	Myanmar	14.697380	
14	Malaysia	13.043224	
15	Nepal	50.437328	
16	Pakistan	31.154530	
17	Philippines (the)	121.748165	
18	Korea (the Democratic People's Republic of)	17.492669	
19	Thailand	22.551331	
20	Taiwan (Province of China)	52.458032	
21	United States of America (the)	78.850049	
22	Viet Nam	0.902428	
23	mean	49.129925	
24	total	1129.988279	

```
CO2_relative_percent_diff_totals CO2_diff_totals tCO2_2015_TRACE \
```

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0	-92.881337	-1.316277e+08	5.861049e+07
1	-174.610414	-2.742306e+07	8.525583e+06
2	-9.887609	-6.605732e+07	1.533412e+08
3	76.811284	5.173455e+06	2.853661e+05
4	51.936188	1.543040e+08	3.209122e+07
5	-38.217479	-2.201589e+08	1.554972e+08
6	12.436318	1.631548e+06	2.193602e+06
7	57.238482	8.169930e+06	1.248992e+06
8	31.134372	1.263498e+07	5.763662e+06
9	-15.406481	-8.925421e+06	1.238675e+07
10	14.655057	2.923897e+06	3.629695e+06
11	84.268427	9.369313e+06	4.152924e+05
12	24.229137	3.343972e+06	2.076407e+06
13	-15.863111	-2.141788e+07	2.830206e+07
14	12.244674	1.884223e+06	2.643498e+06
15	40.279401	7.700079e+06	2.518697e+06
16	-36.903009	-1.835125e+07	1.213108e+07
17	75.679166	1.486006e+08	8.580052e+06
18	-19.169278	-2.001954e+06	2.858041e+06
19	20.266184	4.169196e+07	3.484495e+07
20	-71.109302	-4.341833e+06	1.966739e+06
21	56.553728	2.689073e+07	4.028310e+06
22	-0.906518	-1.532499e+06	3.365033e+07
23	3.599038	-3.370404e+06	2.467779e+07
24	82.777881	-7.751930e+07	5.675891e+08

	tCO2_2016_TRACE	tCO2_2017_TRACE	tCO2_2018_TRACE	tCO2_2019_TRACE \
0	5.695395e+07	5.247394e+07	5.353076e+07	5.177463e+07
1	7.760473e+06	9.312934e+06	9.292575e+06	8.236783e+06
2	1.464883e+08	1.588768e+08	1.353491e+08	1.400838e+08
3	3.337007e+05	3.043248e+05	3.513524e+05	2.870810e+05
4	2.557824e+07	2.403832e+07	2.942454e+07	3.166670e+07
5	1.327353e+08	1.557113e+08	1.647450e+08	1.875389e+08
6	2.295030e+06	2.405054e+06	2.218936e+06	2.375050e+06
7	1.234446e+06	1.360920e+06	1.117475e+06	1.141729e+06
8	5.710333e+06	6.772337e+06	3.870631e+06	5.830141e+06
9	1.432925e+07	1.129261e+07	1.398153e+07	1.486819e+07
10	3.186493e+06	3.658056e+06	3.233858e+06	3.319455e+06
11	4.241102e+05	2.920158e+05	2.524186e+05	3.652645e+05
12	2.529358e+06	1.477960e+06	2.254728e+06	2.119022e+06
13	3.227014e+07	3.012923e+07	3.431117e+07	3.142221e+07
14	2.775123e+06	2.778227e+06	2.666313e+06	2.640717e+06
15	1.666790e+06	2.020325e+06	2.300188e+06	2.910588e+06
16	1.486480e+07	1.343160e+07	1.133074e+07	1.632137e+07
17	1.018389e+07	9.592074e+06	1.043803e+07	8.961374e+06
18	2.294413e+06	2.713641e+06	2.165645e+06	2.413765e+06
19	4.452483e+07	2.911748e+07	2.291644e+07	3.262615e+07
20	2.022287e+06	2.176408e+06	2.034538e+06	2.247717e+06
21	4.046440e+06	4.211999e+06	4.143136e+06	4.228377e+06
22	3.709441e+07	3.516092e+07	3.293637e+07	3.174377e+07
23	2.396966e+07	2.431776e+07	2.368980e+07	2.544012e+07
24	5.513021e+08	5.593084e+08	5.448654e+08	5.851228e+08
	Means	Totals		
0	1.771450e+07	2.733438e+08		
1	1.963150e+06	4.312835e+07		

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2	8.351022e+07	7.341391e+08
3	8.419352e+05	1.561825e+06
4	3.713789e+07	1.427990e+08
5	7.200858e+07	7.962276e+08
6	1.639906e+06	1.148767e+07
7	1.784204e+06	6.103562e+06
8	5.072769e+06	2.794710e+07
9	7.241612e+06	6.685832e+07
10	2.493936e+06	1.702756e+07
11	1.389831e+06	1.749102e+06
12	1.725187e+06	1.045748e+07
13	1.687712e+07	1.564348e+08
14	1.923516e+06	1.350388e+07
15	2.389595e+06	1.141659e+07
16	6.216042e+06	6.807960e+07
17	2.454452e+07	4.775541e+07
18	1.305444e+06	1.244551e+07
19	2.571523e+07	1.640298e+08
20	7.632299e+05	1.044769e+07
21	5.943640e+06	2.065826e+07
22	2.113166e+07	1.705858e+08
23	1.484060e+07	1.220951e+08
24	3.413337e+08	2.808188e+09

---

**Note:** Note a mean of 561,637,600 tonnes CO<sub>2</sub>equiv produced between annually 2015 - 2019.

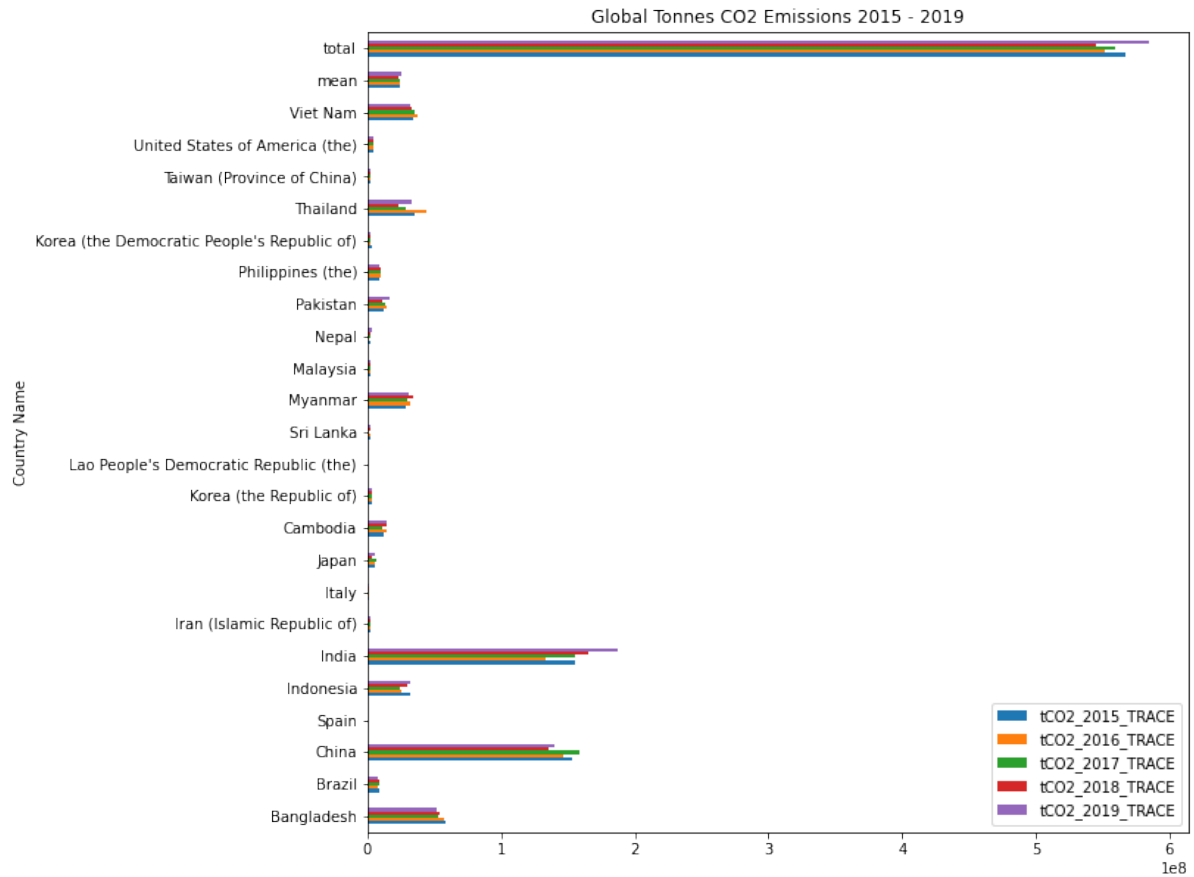
---

## CO<sub>2</sub> Data Insights

- Total difference in 3,370,404 tonnes of CO<sub>2</sub>equiv between the FAOSTAT and TRACE data.
- TRACE reports 2,808,188,000 Tonnes of CO<sub>2</sub>equiv produced between 2015 to 2019.
- There is a relative percent difference of -2.8 percent.

```
TRACE_CO2_df.plot(kind = "barh", x = 'country_name_TRACE', y = ['tCO2_2015_TRACE',
↳ 'tCO2_2016_TRACE', 'tCO2_2017_TRACE', 'tCO2_2018_TRACE', 'tCO2_2019_TRACE'], xlabel =
↳ "Country Name", ylabel = "Tonnes CO2", title = "Global Tonnes CO2 Emissions 2015 -
↳ 2019", figsize = (10,10))
```

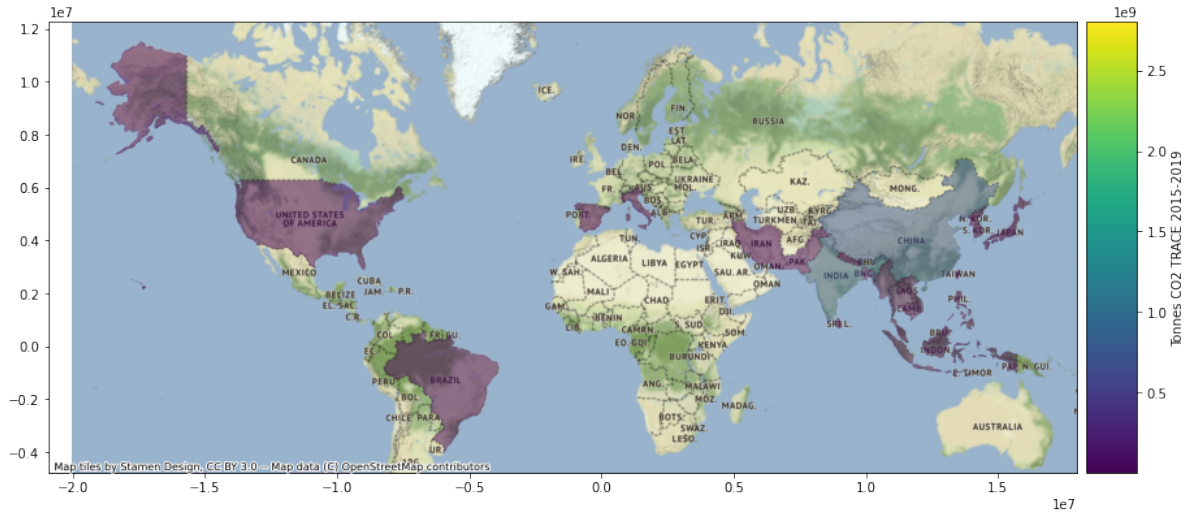
```
<AxesSubplot:title={'center':'Global Tonnes CO2 Emissions 2015 - 2019'}, ylabel=
↳ 'Country Name'>
```



### The Worst Offenders are the Most Dependent on Rice Cultivation

The worst offenders are China, India, and Bangladesh. These states struggle with feeding their massive populations. Including CO2equiv in carbon goals may harm the people of these countries.

```
fig, ax = plt.subplots(1, 1, figsize=(15, 15))
divider = make_axes_locatable(ax)
cax = divider.append_axes("right", size="5%", pad=0.1)
merge_map = merge_geo_df.plot(column='Total_CO2_TRACE', ax=ax, legend=True, cax=cax,
    alpha = .5, legend_kws={'label': "Tonnes CO2 TRACE 2015-2019",
        'orientation': "vertical"})
cx.add_basemap(merge_map, zoom=3)
```



**Note:** Values in billions.

## 2.5 Conclusions

### 2.5.1 FAOSTAT Likely Undercounts Emissions

The TRACE data set on average reports about 2.8 percent more methane emissions than the FAOSTAT data set. This equates to a total methane emissions estimates difference of 3,100,772 tonnes over a 5 year period. The CO<sub>2</sub>equiv over that same period is 77,519,300 tonnes. Neither value is negligible.

### 2.5.2 Countries Should Attempt to Reduce Methane Emissions

The 20 and 100 year warming potentials for methane emission are 84 and 28 (WG1AR5\_Chapter08, Pg 73). These factors will impact our climate if nothing is done. Unfortunately, the countries that produce the most methane from rice cultivation are also those most dependent on the grain. To make the change, realistic solutions must be provided to feed people culturally and geographically dependent on rice cultivation.



## **Part II**

# **Analysis of The University of Malaysia Paper**





## REPLICATING THE UNIVERSITY OF MALAYSIA PAPER.

### 3.1 Imports

```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import scipy.stats as stats
```

### 3.2 University of Malaysia Data

```
filepath = "/Users/jnapolitano/Projects/wattime-takehome/data/ch4_2015-2021.xlsx"

malaysia_emissions_df = pd.read_excel(filepath)
```

```
# Selecting 2020 columns
malaysia_2020 = malaysia_emissions_df[["iso3_country", "country_name", "tCH4_2020"]].
    copy()
```

```
malaysia_2020
```

	iso3_country	country_name	tCH4_2020
0	BGD	Bangladesh	2.106781e+06
1	BRA	Brazil	4.902874e+05
2	CHN	China	6.402353e+06
3	ESP	Spain	1.305461e+04
4	IDN	Indonesia	1.188195e+06
5	IND	India	7.599764e+06
6	IRN	Iran (Islamic Republic of)	9.600254e+04
7	ITA	Italy	5.101547e+04
8	JPN	Japan	2.835167e+05
9	KHM	Cambodia	6.412802e+05
10	KOR	Korea (the Republic of)	1.165467e+05
11	LAO	Lao People's Democratic Republic (the)	2.136270e+04
12	LKA	Sri Lanka	9.248238e+04
13	MMR	Myanmar	1.221904e+06
14	MYS	Malaysia	1.127141e+05
15	NPL	Nepal	7.168401e+04

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16	PAK	Pakistan	6.401201e+05
17	PHL	Philippines (the)	4.462836e+05
18	PRK	Korea (the Democratic People's Republic of)	8.581038e+04
19	THA	Thailand	1.520788e+06
20	TWN	Taiwan (Province of China)	8.333327e+04
21	USA	United States of America (the)	1.941455e+05
22	VNM	Viet Nam	1.374450e+06
23	NaN	Total	2.485387e+07

### 3.3 FAOSTAT Data 2019

```
filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/emissions_csv_fao_emiss_
→ csv_ch4_fao_2015_2019_tonnes.xlsx"
```

```
faostat_emissions_df = pd.read_excel(filepath)
```

```
faostat_2019 = faostat_emissions_df[['code', 'country', 'country_fao', 2019]].copy()
```

```
faostat_2019
```

	code	country \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand
20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2019
0	Bangladesh	1144745.4
1	Brazil	111084.8
2	China, mainland	5214454.7
3	Spain	52098.5

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4	Indonesia	2257604.3
5	India	4621416.8
6	Iran (Islamic Republic of)	96103.4
7	Italy	110895.1
8	Japan	320581.8
9	Cambodia	468378.9
10	Republic of Korea	153260.9
11	Lao People's Democratic Republic	77005.5
12	Sri Lanka	102156.3
13	Myanmar	1083100.3
14	Malaysia	122453.8
15	Nepal	156215.4
16	Pakistan	424755.1
17	Philippines	1556225.8
18	Democratic People's Republic of Korea	82937.0
19	Thailand	1553835.5
20	China, Taiwan Province of	49152.0
21	United States of America	350136.5
22	Viet Nam	1318431.1
23	Total	23446028.9

```
faostat_2019.rename(columns={"code": "iso3_country"}, inplace = True)
```

### 3.4 Merging Data on Iso3 Country Code

```
merged2020_df = faostat_2019.merge(malaysia_2020, on='iso3_country', how='left',
↳sort=False)
```

```
merged2020_df
```

	iso3_country	country \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand

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20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2019 \
0	Bangladesh	1144745.4
1	Brazil	111084.8
2	China, mainland	5214454.7
3	Spain	52098.5
4	Indonesia	2257604.3
5	India	4621416.8
6	Iran (Islamic Republic of)	96103.4
7	Italy	110895.1
8	Japan	320581.8
9	Cambodia	468378.9
10	Republic of Korea	153260.9
11	Lao People's Democratic Republic	77005.5
12	Sri Lanka	102156.3
13	Myanmar	1083100.3
14	Malaysia	122453.8
15	Nepal	156215.4
16	Pakistan	424755.1
17	Philippines	1556225.8
18	Democratic People's Republic of Korea	82937.0
19	Thailand	1553835.5
20	China, Taiwan Province of	49152.0
21	United States of America	350136.5
22	Viet Nam	1318431.1
23	Total	23446028.9

	country_name	tCH4_2020
0	Bangladesh	2.106781e+06
1	Brazil	4.902874e+05
2	China	6.402353e+06
3	Spain	1.305461e+04
4	Indonesia	1.188195e+06
5	India	7.599764e+06
6	Iran (Islamic Republic of)	9.600254e+04
7	Italy	5.101547e+04
8	Japan	2.835167e+05
9	Cambodia	6.412802e+05
10	Korea (the Republic of)	1.165467e+05
11	Lao People's Democratic Republic (the)	2.136270e+04
12	Sri Lanka	9.248238e+04
13	Myanmar	1.221904e+06
14	Malaysia	1.127141e+05
15	Nepal	7.168401e+04
16	Pakistan	6.401201e+05
17	Philippines (the)	4.462836e+05
18	Korea (the Democratic People's Republic of)	8.581038e+04
19	Thailand	1.520788e+06
20	Taiwan (Province of China)	8.333327e+04
21	United States of America (the)	1.941455e+05
22	Viet Nam	1.374450e+06
23	Total	2.485387e+07

### 3.4.1 Calculate difference in Tonnes Between the Estimates

```
# Calculate Difference in tons
merged2020_df['diff_2020'] = merged2020_df[2019] - merged2020_df['tCH4_2020']
```

### 3.4.2 Calculating the Percent Differences Between the Estimates

```
merged2020_df['abs_percent_diff_2020'] = (abs((merged2020_df[2019] - merged2020_df[
↪ 'tCH4_2020']))) / (merged2020_df[2019] + merged2020_df['tCH4_2020']) / 2 * 100
```

```
merged2020_df['relative_percent_diff_2020'] = (merged2020_df[2019] - merged2020_df[
↪ 'tCH4_2020']) / (merged2020_df[2019]) * 100
```

```
merged2020_df
```

	iso3_country	country \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand
20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2019 \
0	Bangladesh	1144745.4
1	Brazil	111084.8
2	China, mainland	5214454.7
3	Spain	52098.5
4	Indonesia	2257604.3
5	India	4621416.8
6	Iran (Islamic Republic of)	96103.4
7	Italy	110895.1
8	Japan	320581.8
9	Cambodia	468378.9
10	Republic of Korea	153260.9

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11	Lao People's Democratic Republic	77005.5
12	Sri Lanka	102156.3
13	Myanmar	1083100.3
14	Malaysia	122453.8
15	Nepal	156215.4
16	Pakistan	424755.1
17	Philippines	1556225.8
18	Democratic People's Republic of Korea	82937.0
19	Thailand	1553835.5
20	China, Taiwan Province of	49152.0
21	United States of America	350136.5
22	Viet Nam	1318431.1
23	Total	23446028.9

	country_name	tCH4_2020	diff_2020	\
0	Bangladesh	2.106781e+06	-9.620360e+05	
1	Brazil	4.902874e+05	-3.792026e+05	
2	China	6.402353e+06	-1.187898e+06	
3	Spain	1.305461e+04	3.904389e+04	
4	Indonesia	1.188195e+06	1.069409e+06	
5	India	7.599764e+06	-2.978347e+06	
6	Iran (Islamic Republic of)	9.600254e+04	1.008569e+02	
7	Italy	5.101547e+04	5.987963e+04	
8	Japan	2.835167e+05	3.706513e+04	
9	Cambodia	6.412802e+05	-1.729013e+05	
10	Korea (the Republic of)	1.165467e+05	3.671416e+04	
11	Lao People's Democratic Republic (the)	2.136270e+04	5.564280e+04	
12	Sri Lanka	9.248238e+04	9.673917e+03	
13	Myanmar	1.221904e+06	-1.388038e+05	
14	Malaysia	1.127141e+05	9.739707e+03	
15	Nepal	7.168401e+04	8.453139e+04	
16	Pakistan	6.401201e+05	-2.153650e+05	
17	Philippines (the)	4.462836e+05	1.109942e+06	
18	Korea (the Democratic People's Republic of)	8.581038e+04	-2.873381e+03	
19	Thailand	1.520788e+06	3.304738e+04	
20	Taiwan (Province of China)	8.333327e+04	-3.418127e+04	
21	United States of America (the)	1.941455e+05	1.559910e+05	
22	Viet Nam	1.374450e+06	-5.601846e+04	
23	Total	2.485387e+07	-1.407846e+06	

	abs_percent_diff_2020	relative_percent_diff_2020
0	14.793604	-84.039294
1	31.528112	-341.363191
2	5.112843	-22.780874
3	29.963187	74.942454
4	15.517575	47.369198
5	12.185184	-64.446615
6	0.026250	0.104946
7	18.491575	53.996641
8	3.067805	11.561831
9	7.790740	-36.914836
10	6.803765	23.955330
11	28.282917	72.258210
12	2.485096	9.469721
13	3.010923	-12.815417
14	2.070799	7.953781

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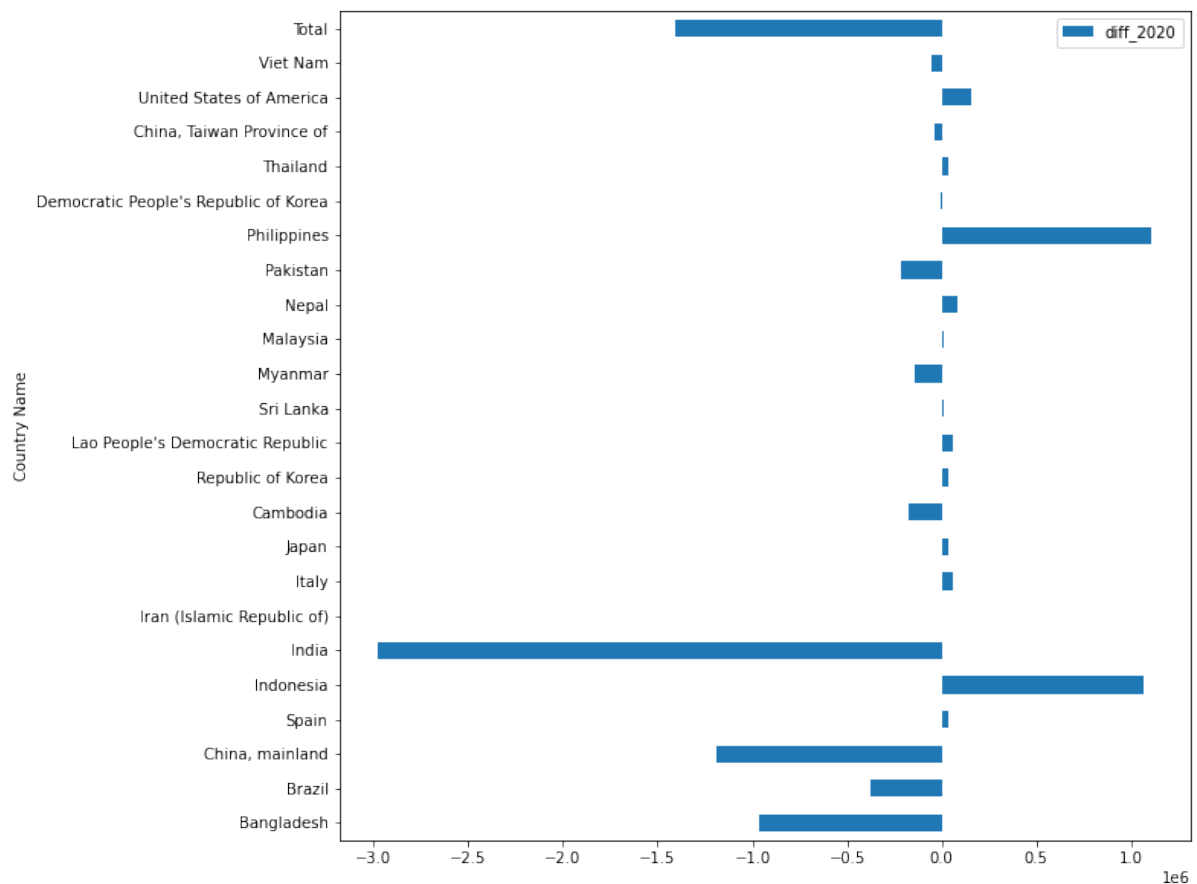
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15	18.545767	54.112071
16	10.112219	-50.703341
17	27.713784	71.322699
18	0.851385	-3.464534
19	0.537422	2.126826
20	12.900027	-69.541975
21	14.329984	44.551492
22	1.040122	-4.248873
23	1.457400	-6.004623

### 3.4.3 Tonnes CH4 FAOSTAT - TRACE Plot

```
merged2020_df.plot(kind = "barh", x = 'country_fao', y = ["diff_2020"], xlabel =
↪ "Country Name", ylabel = "Tonnes CH4", figsize = (10,10))
```

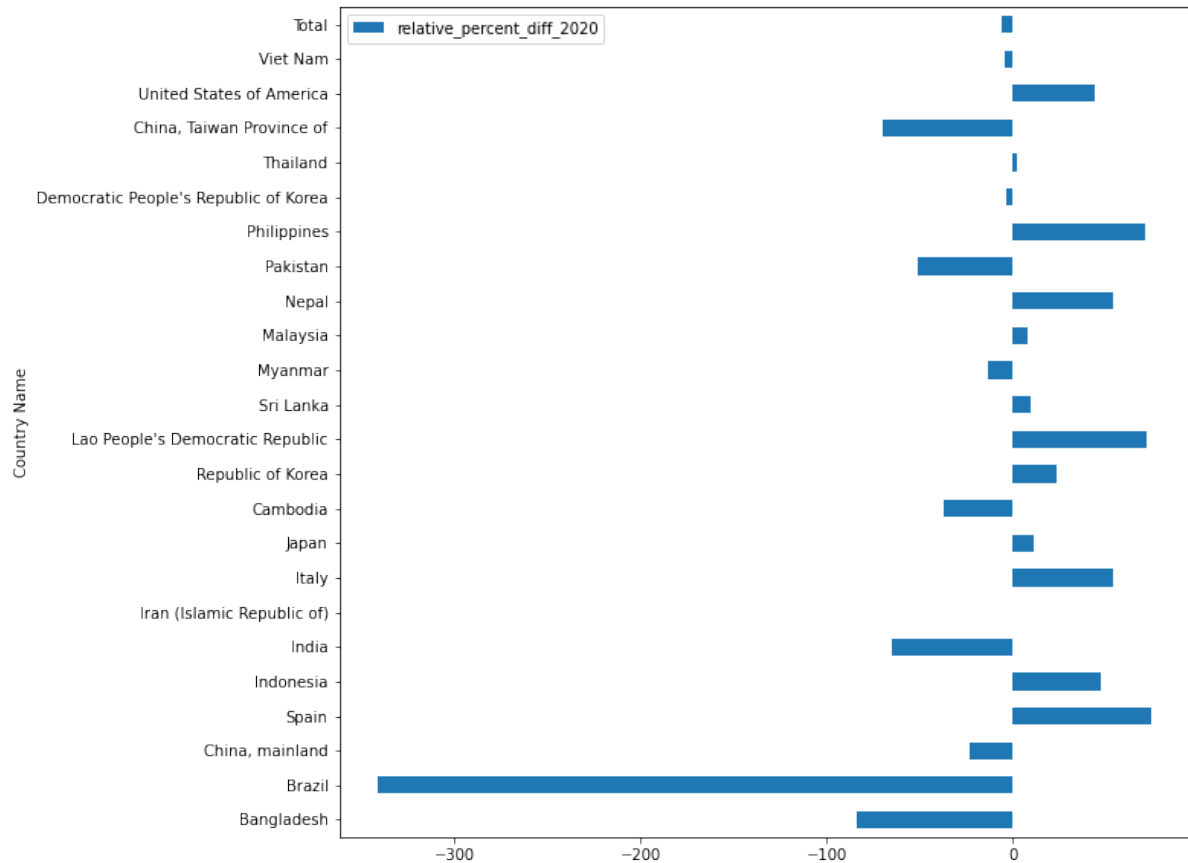
```
<AxesSubplot:ylabel='Country Name'>
```



### 3.4.4 Percent Difference FAOSTAT - TRACE

```
merged2020_df.plot(kind = "barh", x = 'country_fao', y = ["relative_percent_diff_2020",
↪"], xlabel = "Country Name", ylabel = "Tonnes CH4", figsize=(10,10))
```

```
<AxesSubplot:ylabel='Country Name'>
```



## 3.5 Recreating the Malaysia Paper with 2018 Data.

### 3.5.1 University of Malaysia Data

```
filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/ch4_2015-2021.xlsx"
malaysia_emissions_df = pd.read_excel(filepath)
```

```
# Selecting 2020 columns
malaysia_2020 = malaysia_emissions_df[["iso3_country", "country_name", "tCH4_2020"]].
↪copy()
```

```
malaysia_2020
```



	iso3_country	country_name	tCH4_2020
0	BGD	Bangladesh	2.106781e+06
1	BRA	Brazil	4.902874e+05
2	CHN	China	6.402353e+06
3	ESP	Spain	1.305461e+04
4	IDN	Indonesia	1.188195e+06
5	IND	India	7.599764e+06
6	IRN	Iran (Islamic Republic of)	9.600254e+04
7	ITA	Italy	5.101547e+04
8	JPN	Japan	2.835167e+05
9	KHM	Cambodia	6.412802e+05
10	KOR	Korea (the Republic of)	1.165467e+05
11	LAO	Lao People's Democratic Republic (the)	2.136270e+04
12	LKA	Sri Lanka	9.248238e+04
13	MMR	Myanmar	1.221904e+06
14	MYS	Malaysia	1.127141e+05
15	NPL	Nepal	7.168401e+04
16	PAK	Pakistan	6.401201e+05
17	PHL	Philippines (the)	4.462836e+05
18	PRK	Korea (the Democratic People's Republic of)	8.581038e+04
19	THA	Thailand	1.520788e+06
20	TWN	Taiwan (Province of China)	8.333327e+04
21	USA	United States of America (the)	1.941455e+05
22	VNM	Viet Nam	1.374450e+06
23	NaN	Total	2.485387e+07

### 3.6 FAOSTAT Data 2018

```
filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/emissions_csv_fao_emiss_
→ csv_ch4_fao_2015_2019_tonnes.xlsx"
```

```
faostat_emissions_df = pd.read_excel(filepath)
```

```
faostat_2018 = faostat_emissions_df[['code', 'country', 'country_fao', 2018]].copy()
```

```
faostat_2018
```

	code	country \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka

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```

13 MMR Myanmar
14 MYS Malaysia
15 NPL Nepal
16 PAK Pakistan
17 PHL Philippines (the)
18 PRK Korea (the Democratic People's Republic of)
19 THA Thailand
20 TWN Taiwan (Province of China)
21 USA United States of America (the)
22 VNM Viet Nam
23 NaN NaN

```

```

country_fao 2018
0 Bangladesh 1144591.0
1 Brazil 121615.2
2 China, mainland 5302173.1
3 Spain 52925.0
4 Indonesia 2405613.8
5 India 4661154.9
6 Iran (Islamic Republic of) 93936.6
7 Italy 109463.8
8 Japan 322245.0
9 Cambodia 479362.7
10 Republic of Korea 154911.3
11 Lao People's Democratic Republic 83333.6
12 Sri Lanka 111049.0
13 Myanmar 1118850.0
14 Malaysia 125238.5
15 Nepal 153890.8
16 Pakistan 393404.2
17 Philippines 1606047.8
18 Democratic People's Republic of Korea 83943.3
19 Thailand 1702989.1
20 China, Taiwan Province of 49414.1
21 United States of America 412177.5
22 Viet Nam 1336231.2
23 Total 24042561.5

```

```
faostat_2018.rename(columns={"code": "iso3_country"}, inplace = True)
```

### 3.6.1 Merging Data on Iso3 Country Code

```
merged2020_df = faostat_2018.merge(malaysia_2020, on='iso3_country', how='left',
↳sort=False)
```

```
merged2020_df
```

```

iso3_country      country \
0 BGD Bangladesh
1 BRA Brazil
2 CHN China

```

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3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand
20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2018	\
0	Bangladesh	1144591.0	
1	Brazil	121615.2	
2	China, mainland	5302173.1	
3	Spain	52925.0	
4	Indonesia	2405613.8	
5	India	4661154.9	
6	Iran (Islamic Republic of)	93936.6	
7	Italy	109463.8	
8	Japan	322245.0	
9	Cambodia	479362.7	
10	Republic of Korea	154911.3	
11	Lao People's Democratic Republic	83333.6	
12	Sri Lanka	111049.0	
13	Myanmar	1118850.0	
14	Malaysia	125238.5	
15	Nepal	153890.8	
16	Pakistan	393404.2	
17	Philippines	1606047.8	
18	Democratic People's Republic of Korea	83943.3	
19	Thailand	1702989.1	
20	China, Taiwan Province of	49414.1	
21	United States of America	412177.5	
22	Viet Nam	1336231.2	
23	Total	24042561.5	

	country_name	tCH4_2020
0	Bangladesh	2.106781e+06
1	Brazil	4.902874e+05
2	China	6.402353e+06
3	Spain	1.305461e+04
4	Indonesia	1.188195e+06
5	India	7.599764e+06
6	Iran (Islamic Republic of)	9.600254e+04

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7	Italy	5.101547e+04
8	Japan	2.835167e+05
9	Cambodia	6.412802e+05
10	Korea (the Republic of)	1.165467e+05
11	Lao People's Democratic Republic (the)	2.136270e+04
12	Sri Lanka	9.248238e+04
13	Myanmar	1.221904e+06
14	Malaysia	1.127141e+05
15	Nepal	7.168401e+04
16	Pakistan	6.401201e+05
17	Philippines (the)	4.462836e+05
18	Korea (the Democratic People's Republic of)	8.581038e+04
19	Thailand	1.520788e+06
20	Taiwan (Province of China)	8.333327e+04
21	United States of America (the)	1.941455e+05
22	Viet Nam	1.374450e+06
23	Total	2.485387e+07

### 3.6.2 Calculate difference in Tonnes Between the Estimates

```
# Calculate Difference in tons
merged2020_df['diff_2020'] = merged2020_df[2018] - merged2020_df['tCH4_2020']
```

### 3.6.3 Calculating the Percent Differences Between the Estimates

```
merged2020_df['abs_percent_diff_2020'] = (abs((merged2020_df[2018] - merged2020_df[
↪ 'tCH4_2020']))) / (merged2020_df[2018] + merged2020_df["tCH4_2020"]) / 2) * 100

merged2020_df['relative_percent_diff_2020'] = (merged2020_df[2018] - merged2020_df[
↪ 'tCH4_2020']) / (merged2020_df[2018]) * 100
```

```
merged2020_df
```

	iso3_country	country \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal

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16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand
20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2018	\
0	Bangladesh	1144591.0	
1	Brazil	121615.2	
2	China, mainland	5302173.1	
3	Spain	52925.0	
4	Indonesia	2405613.8	
5	India	4661154.9	
6	Iran (Islamic Republic of)	93936.6	
7	Italy	109463.8	
8	Japan	322245.0	
9	Cambodia	479362.7	
10	Republic of Korea	154911.3	
11	Lao People's Democratic Republic	83333.6	
12	Sri Lanka	111049.0	
13	Myanmar	1118850.0	
14	Malaysia	125238.5	
15	Nepal	153890.8	
16	Pakistan	393404.2	
17	Philippines	1606047.8	
18	Democratic People's Republic of Korea	83943.3	
19	Thailand	1702989.1	
20	China, Taiwan Province of	49414.1	
21	United States of America	412177.5	
22	Viet Nam	1336231.2	
23	Total	24042561.5	

	country_name	tCH4_2020	diff_2020	\
0	Bangladesh	2.106781e+06	-9.621904e+05	
1	Brazil	4.902874e+05	-3.686722e+05	
2	China	6.402353e+06	-1.100180e+06	
3	Spain	1.305461e+04	3.987039e+04	
4	Indonesia	1.188195e+06	1.217419e+06	
5	India	7.599764e+06	-2.938609e+06	
6	Iran (Islamic Republic of)	9.600254e+04	-2.065943e+03	
7	Italy	5.101547e+04	5.844833e+04	
8	Japan	2.835167e+05	3.872833e+04	
9	Cambodia	6.412802e+05	-1.619175e+05	
10	Korea (the Republic of)	1.165467e+05	3.836456e+04	
11	Lao People's Democratic Republic (the)	2.136270e+04	6.197090e+04	
12	Sri Lanka	9.248238e+04	1.856662e+04	
13	Myanmar	1.221904e+06	-1.030541e+05	
14	Malaysia	1.127141e+05	1.252441e+04	
15	Nepal	7.168401e+04	8.220679e+04	
16	Pakistan	6.401201e+05	-2.467159e+05	
17	Philippines (the)	4.462836e+05	1.159764e+06	
18	Korea (the Democratic People's Republic of)	8.581038e+04	-1.867081e+03	
19	Thailand	1.520788e+06	1.822010e+05	

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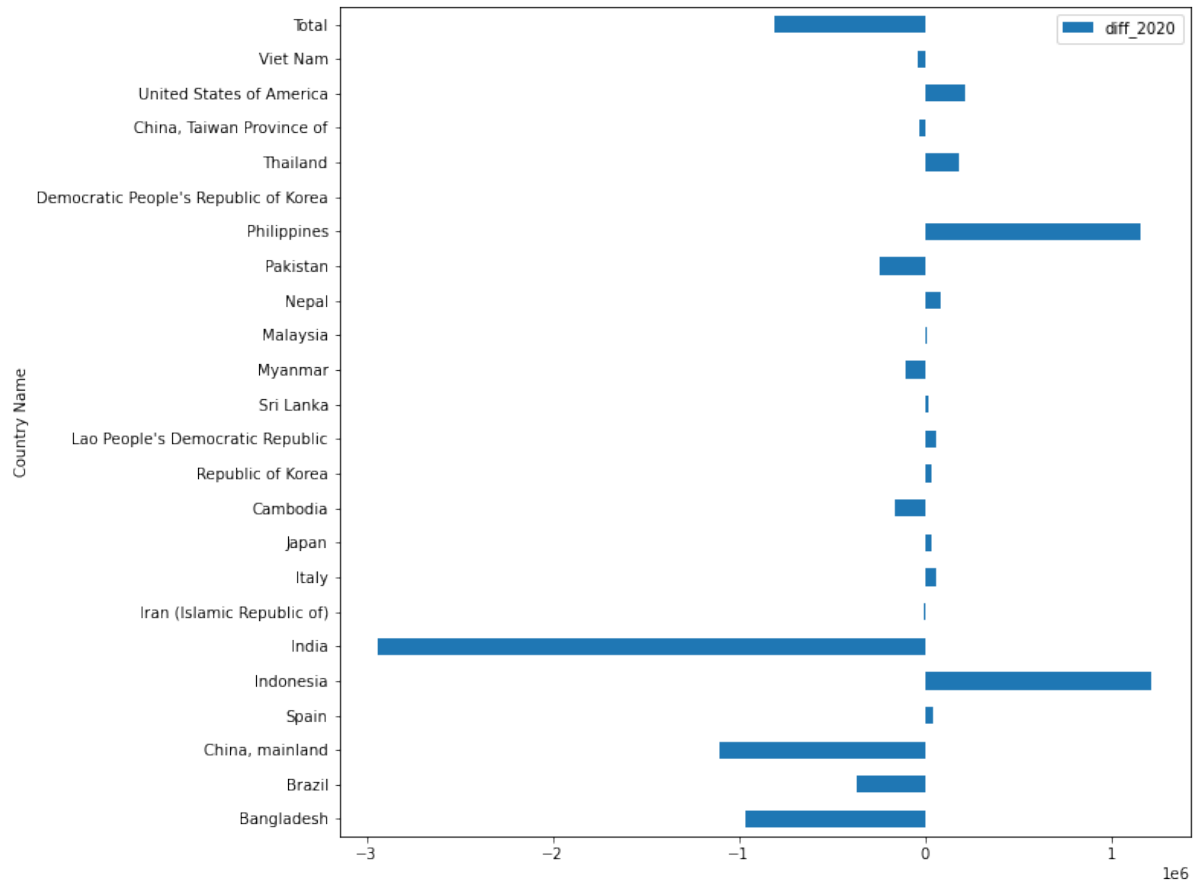
20	Taiwan (Province of China)	8.333327e+04	-3.391917e+04
21	United States of America (the)	1.941455e+05	2.180320e+05
22	Viet Nam	1.374450e+06	-3.821836e+04
23	Total	2.485387e+07	-8.113131e+05

	abs_percent_diff_2020	relative_percent_diff_2020
0	14.796680	-84.064120
1	30.125073	-303.146497
2	4.699806	-20.749605
3	30.214181	75.333763
4	16.937719	50.607398
5	11.983640	-63.044646
6	0.543843	-2.199295
7	18.210554	53.395122
8	3.196664	12.018286
9	7.224313	-33.777660
10	7.066388	24.765498
11	29.595551	74.364837
12	4.561119	16.719301
13	2.201302	-9.210718
14	2.631702	10.000444
15	18.221624	53.418910
16	11.935661	-62.713090
17	28.254800	72.212312
18	0.549938	-2.224216
19	2.825893	10.698893
20	12.775835	-68.642698
21	17.979860	52.897608
22	0.704959	-2.860161
23	0.829624	-3.374487

### 3.6.4 Tonnes CH4 FAOSTAT - TRACE Plot

```
merged2020_df.plot(kind = "barh", x = 'country_fao', y = ["diff_2020"], xlabel =
    "Country Name", ylabel = "Tonnes CH4", figsize = (10,10))
```

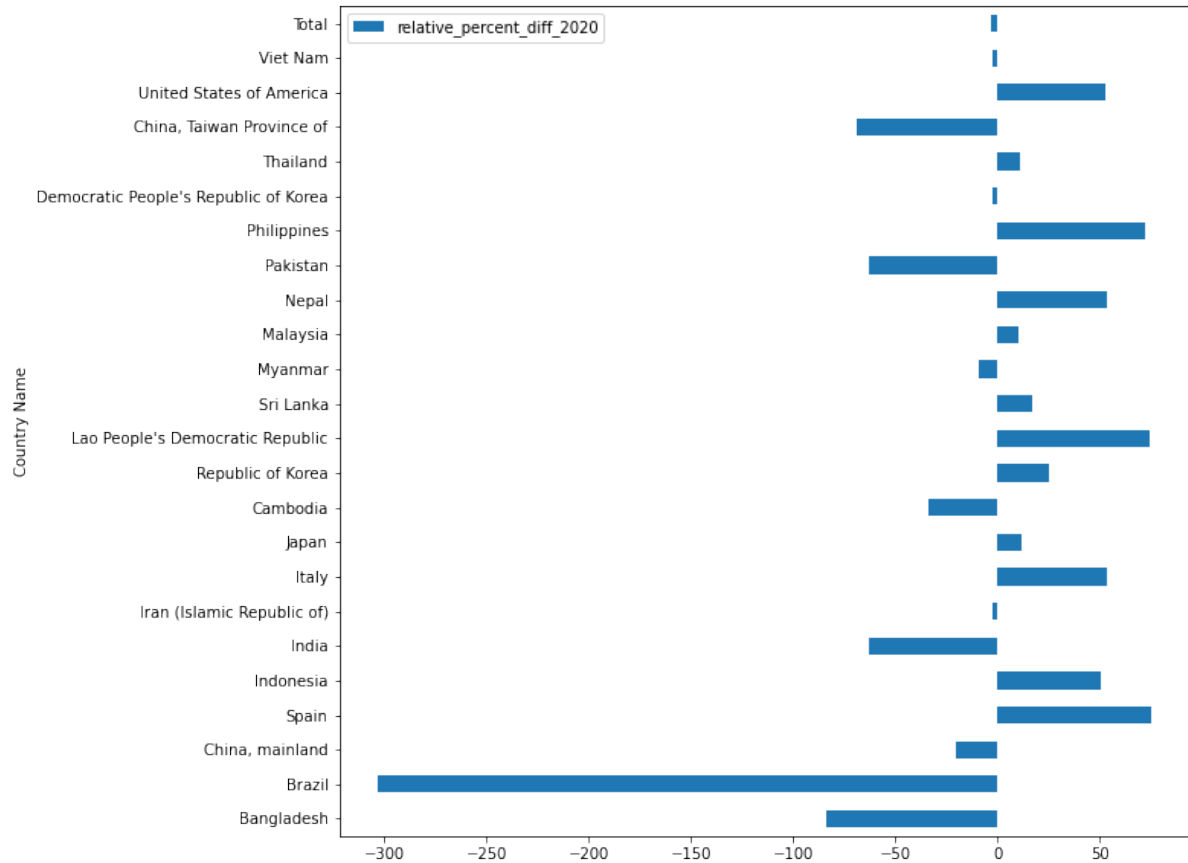
```
<AxesSubplot:ylabel='Country Name'>
```



### 3.6.5 Percent Difference FAOSTAT - TRACE

```
merged2020_df.plot(kind = "barh", x = 'country_fao', y = ["relative_percent_diff_2020", "↔"], xlabel = "Country Name", ylabel = "Tonnes CH4", figsize=(10,10))
```

```
<AxesSubplot:ylabel='Country Name'>
```



## 3.7 Recreating the Malaysia Paper same year

### 3.7.1 University of Malaysia Data

```
filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/ch4_2015-2021.xlsx"
```

```
malaysia_emissions_df = pd.read_excel(filepath)
```

```
# Selecting 2020 columns
```

```
malaysia_2018 = malaysia_emissions_df[["iso3_country", "country_name", "tCH4_2018"]].  
↪copy()
```

```
malaysia_2018
```

	iso3_country	country_name	tCH4_2018
0	BGD	Bangladesh	2.141231e+06
1	BRA	Brazil	3.717030e+05
2	CHN	China	5.413962e+06
3	ESP	Spain	1.405410e+04
4	IDN	Indonesia	1.176982e+06
5	IND	India	6.589798e+06

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6	IRN	Iran (Islamic Republic of)	8.875744e+04
7	ITA	Italy	4.469902e+04
8	JPN	Japan	1.548252e+05
9	KHM	Cambodia	5.592610e+05
10	KOR	Korea (the Republic of)	1.293543e+05
11	LAO	Lao People's Democratic Republic (the)	1.009675e+04
12	LKA	Sri Lanka	9.018914e+04
13	MMR	Myanmar	1.372447e+06
14	MYS	Malaysia	1.066525e+05
15	NPL	Nepal	9.200752e+04
16	PAK	Pakistan	4.532297e+05
17	PHL	Philippines (the)	4.175210e+05
18	PRK	Korea (the Democratic People's Republic of)	8.662578e+04
19	THA	Thailand	9.166575e+05
20	TWN	Taiwan (Province of China)	8.138151e+04
21	USA	United States of America (the)	1.657254e+05
22	VNM	Viet Nam	1.317455e+06
23	NaN	Total	2.179462e+07

### 3.8 FAOSTAT Data 2018

```
filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/emissions_csv_fao_emiss-
csv_ch4_fao_2015_2019_tonnes.xlsx"
```

```
faostat_emissions_df = pd.read_excel(filepath)
```

```
faostat_2018 = faostat_emissions_df[['code', 'country', 'country_fao', 2018]].copy()
```

```
faostat_2018
```

	code	country \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand

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```

20 TWN                Taiwan (Province of China)
21 USA                United States of America (the)
22 VNM                Viet Nam
23 NaN                NaN

                                country_fao      2018
0                                Bangladesh  1144591.0
1                                Brazil      121615.2
2                                China, mainland  5302173.1
3                                Spain        52925.0
4                                Indonesia   2405613.8
5                                India       4661154.9
6                                Iran (Islamic Republic of)  93936.6
7                                Italy       109463.8
8                                Japan       322245.0
9                                Cambodia   479362.7
10                               Republic of Korea  154911.3
11                               Lao People's Democratic Republic  83333.6
12                               Sri Lanka    111049.0
13                               Myanmar    1118850.0
14                               Malaysia    125238.5
15                               Nepal       153890.8
16                               Pakistan    393404.2
17                               Philippines  1606047.8
18                               Democratic People's Republic of Korea  83943.3
19                               Thailand    1702989.1
20                               China, Taiwan Province of  49414.1
21                               United States of America  412177.5
22                               Viet Nam    1336231.2
23                               Total      24042561.5

```

```
faostat_2018.rename(columns={"code": "iso3_country"}, inplace = True)
```

### 3.8.1 Merging Data on Iso3 Country Code

```
merged2018_df = faostat_2018.merge(malaysia_2018, on='iso3_country', how='left',
↳sort=False)
```

```
merged2018_df
```

```

iso3_country      country \
0      BGD      Bangladesh
1      BRA      Brazil
2      CHN      China
3      ESP      Spain
4      IDN      Indonesia
5      IND      India
6      IRN      Iran (Islamic Republic of)
7      ITA      Italy
8      JPN      Japan
9      KHM      Cambodia

```

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10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand
20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2018 \
0	Bangladesh	1144591.0
1	Brazil	121615.2
2	China, mainland	5302173.1
3	Spain	52925.0
4	Indonesia	2405613.8
5	India	4661154.9
6	Iran (Islamic Republic of)	93936.6
7	Italy	109463.8
8	Japan	322245.0
9	Cambodia	479362.7
10	Republic of Korea	154911.3
11	Lao People's Democratic Republic	83333.6
12	Sri Lanka	111049.0
13	Myanmar	1118850.0
14	Malaysia	125238.5
15	Nepal	153890.8
16	Pakistan	393404.2
17	Philippines	1606047.8
18	Democratic People's Republic of Korea	83943.3
19	Thailand	1702989.1
20	China, Taiwan Province of	49414.1
21	United States of America	412177.5
22	Viet Nam	1336231.2
23	Total	24042561.5

	country_name	tCH4_2018
0	Bangladesh	2.141231e+06
1	Brazil	3.717030e+05
2	China	5.413962e+06
3	Spain	1.405410e+04
4	Indonesia	1.176982e+06
5	India	6.589798e+06
6	Iran (Islamic Republic of)	8.875744e+04
7	Italy	4.469902e+04
8	Japan	1.548252e+05
9	Cambodia	5.592610e+05
10	Korea (the Republic of)	1.293543e+05
11	Lao People's Democratic Republic (the)	1.009675e+04
12	Sri Lanka	9.018914e+04
13	Myanmar	1.372447e+06

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```

14                                     Malaysia  1.066525e+05
15                                     Nepal    9.200752e+04
16                                     Pakistan  4.532297e+05
17                                     Philippines (the) 4.175210e+05
18 Korea (the Democratic People's Republic of) 8.662578e+04
19                                     Thailand  9.166575e+05
20                                     Taiwan (Province of China) 8.138151e+04
21 United States of America (the) 1.657254e+05
22 Viet Nam 1.317455e+06
23 Total 2.179462e+07

```

### 3.8.2 Calculate difference in Tonnes Between the Estimates

```

# Calculate Difference in tons
merged2018_df['diff_2018'] = merged2018_df[2018] - merged2018_df['tCH4_2018']

```

### 3.8.3 Calculating the Percent Differences Between the Estimates

```

merged2018_df['abs_percent_diff_2018'] = (abs((merged2018_df[2018] - merged2018_df[
↪ 'tCH4_2018']))) / (merged2018_df[2018] + merged2018_df["tCH4_2018"]) / 2 * 100

merged2018_df['relative_percent_diff_2018'] = (merged2018_df[2018] - merged2018_df[
↪ 'tCH4_2018']) / (merged2018_df[2018]) * 100

```

```
merged2018_df
```

```

iso3_country      country \
0      BGD      Bangladesh
1      BRA      Brazil
2      CHN      China
3      ESP      Spain
4      IDN      Indonesia
5      IND      India
6      IRN      Iran (Islamic Republic of)
7      ITA      Italy
8      JPN      Japan
9      KHM      Cambodia
10     KOR      Korea (the Republic of)
11     LAO      Lao People's Democratic Republic (the)
12     LKA      Sri Lanka
13     MMR      Myanmar
14     MYS      Malaysia
15     NPL      Nepal
16     PAK      Pakistan
17     PHL      Philippines (the)
18     PRK      Korea (the Democratic People's Republic of)
19     THA      Thailand
20     TWN      Taiwan (Province of China)
21     USA      United States of America (the)
22     VNM      Viet Nam

```

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23	NaN	NaN
	country_fao	2018 \
0	Bangladesh	1144591.0
1	Brazil	121615.2
2	China, mainland	5302173.1
3	Spain	52925.0
4	Indonesia	2405613.8
5	India	4661154.9
6	Iran (Islamic Republic of)	93936.6
7	Italy	109463.8
8	Japan	322245.0
9	Cambodia	479362.7
10	Republic of Korea	154911.3
11	Lao People's Democratic Republic	83333.6
12	Sri Lanka	111049.0
13	Myanmar	1118850.0
14	Malaysia	125238.5
15	Nepal	153890.8
16	Pakistan	393404.2
17	Philippines	1606047.8
18	Democratic People's Republic of Korea	83943.3
19	Thailand	1702989.1
20	China, Taiwan Province of	49414.1
21	United States of America	412177.5
22	Viet Nam	1336231.2
23	Total	24042561.5
	country_name	tCH4_2018 diff_2018 \
0	Bangladesh	2.141231e+06 -9.966395e+05
1	Brazil	3.717030e+05 -2.500878e+05
2	China	5.413962e+06 -1.117892e+05
3	Spain	1.405410e+04 3.887090e+04
4	Indonesia	1.176982e+06 1.228632e+06
5	India	6.589798e+06 -1.928643e+06
6	Iran (Islamic Republic of)	8.875744e+04 5.179159e+03
7	Italy	4.469902e+04 6.476478e+04
8	Japan	1.548252e+05 1.674198e+05
9	Cambodia	5.592610e+05 -7.989830e+04
10	Korea (the Republic of)	1.293543e+05 2.555698e+04
11	Lao People's Democratic Republic (the)	1.009675e+04 7.323685e+04
12	Sri Lanka	9.018914e+04 2.085986e+04
13	Myanmar	1.372447e+06 -2.535969e+05
14	Malaysia	1.066525e+05 1.858597e+04
15	Nepal	9.200752e+04 6.188328e+04
16	Pakistan	4.532297e+05 -5.982552e+04
17	Philippines (the)	4.175210e+05 1.188527e+06
18	Korea (the Democratic People's Republic of)	8.662578e+04 -2.682481e+03
19	Thailand	9.166575e+05 7.863316e+05
20	Taiwan (Province of China)	8.138151e+04 -3.196741e+04
21	United States of America (the)	1.657254e+05 2.464521e+05
22	Viet Nam	1.317455e+06 1.877621e+04
23	Total	2.179462e+07 2.247946e+06
	abs_percent_diff_2018	relative_percent_diff_2018
0	15.165759	-87.073857

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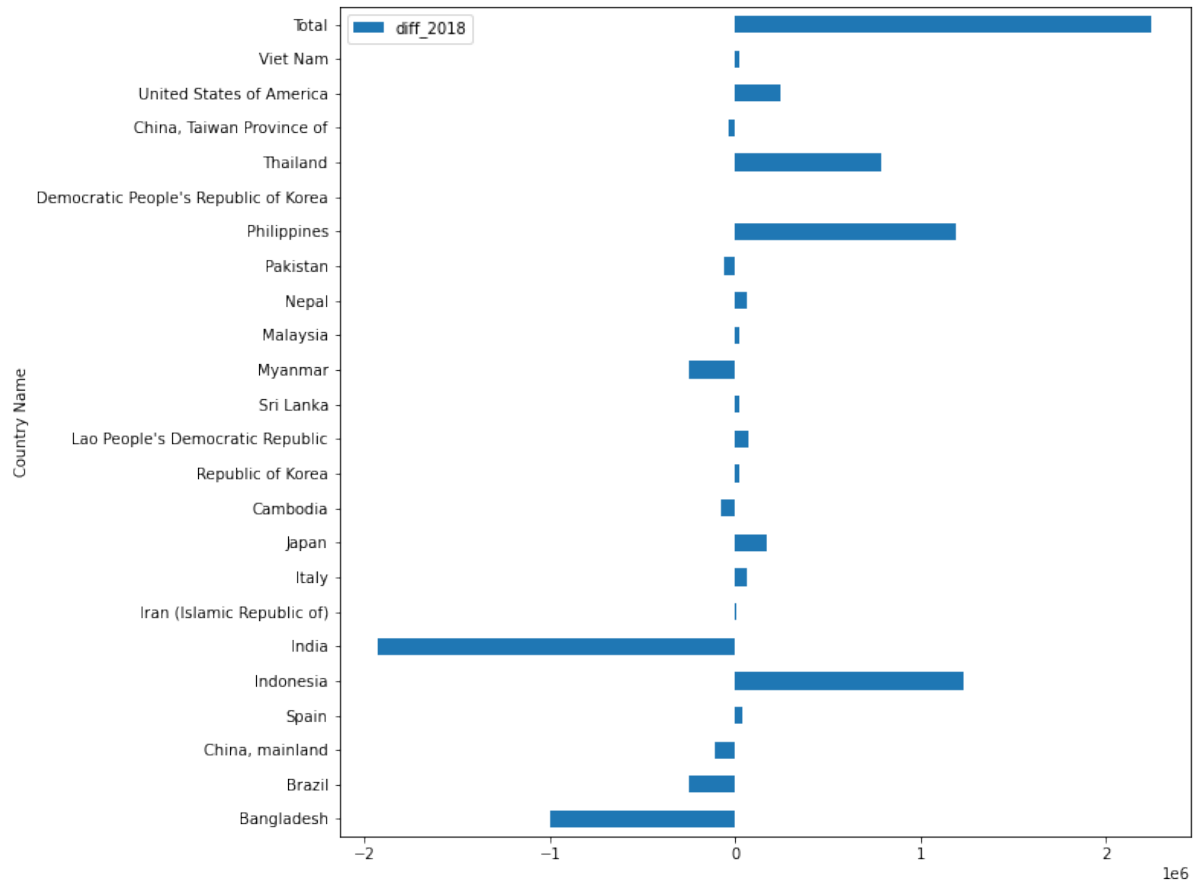
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1	25.347513	-205.638590
2	0.521593	-2.108366
3	29.017189	73.445258
4	17.147238	51.073546
5	8.571022	-41.376939
6	1.417441	5.513463
7	21.005319	59.165480
8	17.546657	51.954183
9	3.846355	-16.667610
10	4.495264	16.497815
11	39.193291	87.883944
12	5.182880	18.784377
13	5.089656	-22.665852
14	4.007480	14.840462
15	12.583102	40.212459
16	3.533140	-15.207139
17	29.367095	74.003201
18	0.786333	-3.195586
19	15.008352	46.173612
20	12.220368	-64.692890
21	21.322962	59.792699
22	0.353776	1.405162
23	2.452099	9.349861

### 3.8.4 Tonnes CH4 FAOSTAT - TRACE Plot

```
merged2018_df.plot(kind = "barh", x = 'country_fao', y = ["diff_2018"], xlabel =
↪ "Country Name", ylabel = "Tonnes CH4", figsize = (10,10))
```

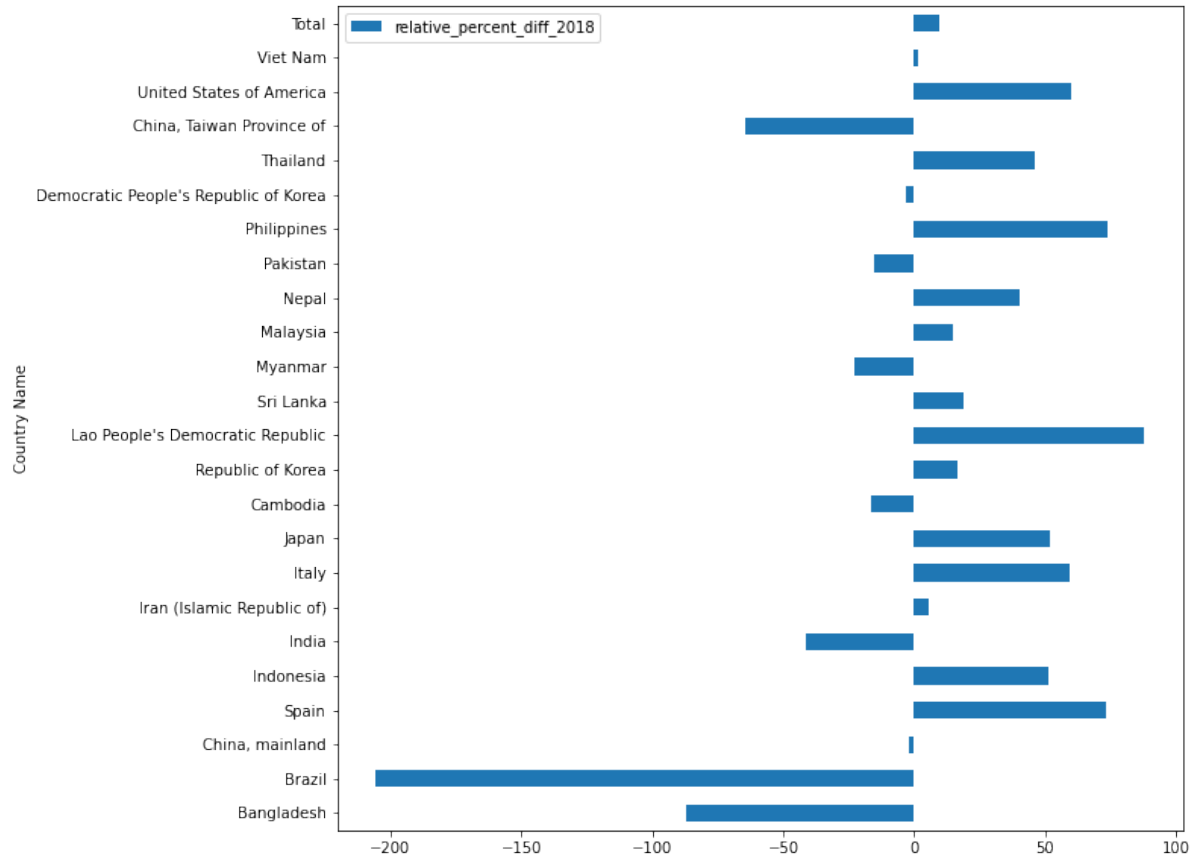
```
<AxesSubplot:ylabel='Country Name'>
```



### 3.8.5 Percent Difference FAOSTAT - TRACE

```
merged2018_df.plot(kind = "barh", x = 'country_fao', y = ["relative_percent_diff_2018", "↔"], xlabel = "Country Name", ylabel = "Tonnes CH4", figsize=(10,10))
```

```
<AxesSubplot:ylabel='Country Name'>
```



### 3.9 Impressions

I was not able to replicate the results exactly. I reviewed the included paper to find that the university misleadingly rounded to three decimal places. I did not manipulate the raw data.

I found that the overall difference in reporting is only about 6 percent between 2019 and 2020. The methodology is somewhat misleading as well, because when considering the previous 5 years of data demonstrated a consistent 3 percent difference. When comparing the TRACE data to itself it does seem to show a significant increase between 2019 and 2020 emissions. I'll test this next.



## HYPOTHESIS TESTING THE UNIVERSITY OF MALAYSIA PAPER

### 4.1 Claims

- That the distributions do not differ between 2020 and 2019
- That the means do not differ between 2020 and 2019

### 4.2 What will be testing.

- That the Data are independent and evenly distributed: Test for normality
  - Shapiro-Wilk Test
- That the means between 2019 and 2020 do not differ: Parametric Statistical Hypothesis Tests
  - T Test because we have less than 25 observations
- If nonparametric: That the distributions between 2019 and 2020 do not differ Mann-Whitney U Test

### 4.3 Data Import

```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import scipy.stats as stats
```

```
filepath = "/Users/jnapolitano/Projects/wattime-takehome/data/ch4_2015-2021.xlsx"

hypothesis_testing_df = pd.read_excel(filepath)
```

### 4.3.1 Drop total row from the data

```
hypothesis_testing_df = hypothesis_testing_df.loc[(hypothesis_testing_df['country_name'
↪'] != "Total")].copy() #copying to avoid modifying slices in memory. Old df should
↪also drop from memory in production environment.
```

```
hypothesis_testing_df
```

	iso3_country	country_name	tCH4_2015	\		
0	BGD	Bangladesh	2.344420e+06			
1	BRA	Brazil	3.410233e+05			
2	CHN	China	6.133647e+06			
3	ESP	Spain	1.141464e+04			
4	IDN	Indonesia	1.283649e+06			
5	IND	India	6.219887e+06			
6	IRN	Iran (Islamic Republic of)	8.774407e+04			
7	ITA	Italy	4.995968e+04			
8	JPN	Japan	2.305465e+05			
9	KHM	Cambodia	4.954698e+05			
10	KOR	Korea (the Republic of)	1.451878e+05			
11	LAO	Lao People's Democratic Republic (the)	1.661169e+04			
12	LKA	Sri Lanka	8.305626e+04			
13	MMR	Myanmar	1.132082e+06			
14	MYS	Malaysia	1.057399e+05			
15	NPL	Nepal	1.007479e+05			
16	PAK	Pakistan	4.852431e+05			
17	PHL	Philippines (the)	3.432021e+05			
18	PRK	Korea (the Democratic People's Republic of)	1.143217e+05			
19	THA	Thailand	1.393798e+06			
20	TWN	Taiwan (Province of China)	7.866956e+04			
21	USA	United States of America (the)	1.611324e+05			
22	VNM	Viet Nam	1.346013e+06			
	tCH4_2016	tCH4_2017	tCH4_2018	tCH4_2019	tCH4_2020	\
0	2.278158e+06	2.098958e+06	2.141231e+06	2.070985e+06	2.106781e+06	
1	3.104189e+05	3.725173e+05	3.717030e+05	3.294713e+05	4.902874e+05	
2	5.859531e+06	6.355071e+06	5.413962e+06	5.603352e+06	6.402353e+06	
3	1.334803e+04	1.217299e+04	1.405410e+04	1.148324e+04	1.305461e+04	
4	1.023129e+06	9.615327e+05	1.176982e+06	1.266668e+06	1.188195e+06	
5	5.309413e+06	6.228451e+06	6.589798e+06	7.501556e+06	7.599764e+06	
6	9.180121e+04	9.620217e+04	8.875744e+04	9.500199e+04	9.600254e+04	
7	4.937785e+04	5.443679e+04	4.469902e+04	4.566914e+04	5.101547e+04	
8	2.284133e+05	2.708935e+05	1.548252e+05	2.332056e+05	2.835167e+05	
9	5.731698e+05	4.517045e+05	5.592610e+05	5.947277e+05	6.412802e+05	
10	1.274597e+05	1.463222e+05	1.293543e+05	1.327782e+05	1.165467e+05	
11	1.696441e+04	1.168063e+04	1.009675e+04	1.461058e+04	2.136270e+04	
12	1.011743e+05	5.911841e+04	9.018914e+04	8.476088e+04	9.248238e+04	
13	1.290806e+06	1.205169e+06	1.372447e+06	1.256888e+06	1.221904e+06	
14	1.110049e+05	1.111291e+05	1.066525e+05	1.056287e+05	1.127141e+05	
15	6.667161e+04	8.081300e+04	9.200752e+04	1.164235e+05	7.168401e+04	
16	5.945922e+05	5.372641e+05	4.532297e+05	6.528548e+05	6.401201e+05	
17	4.073554e+05	3.836830e+05	4.175210e+05	3.584550e+05	4.462836e+05	
18	9.177653e+04	1.085457e+05	8.662578e+04	9.655062e+04	8.581038e+04	
19	1.780993e+06	1.164699e+06	9.166575e+05	1.305046e+06	1.520788e+06	
20	8.089149e+04	8.705634e+04	8.138151e+04	8.990870e+04	8.333327e+04	
21	1.618576e+05	1.684799e+05	1.657254e+05	1.691351e+05	1.941455e+05	

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```

22  1.483777e+06  1.406437e+06  1.317455e+06  1.269751e+06  1.374450e+06

      tCH4_2021
0    1.983974e+06
1    4.544874e+05
2    6.068210e+06
3    8.531579e+03
4    1.009936e+06
5    6.567960e+06
6    9.053525e+04
7    5.089759e+04
8    1.574007e+05
9    5.644891e+05
10   1.013006e+05
11   1.475014e+04
12   8.466966e+04
13   1.289837e+06
14   1.069696e+05
15   4.811408e+04
16   4.849205e+05
17   4.383270e+05
18   7.735988e+04
19   8.528673e+05
20   6.619861e+04
21   1.634842e+05
22   1.502787e+06

```

### 4.3.2 Test for Normality: Shapiro-Wilk

2019

```

## Selecting Malaysia 2019 Data
data_2019 = hypothesis_testing_df['tCH4_2019']
data_2019

```

```

0    2.070985e+06
1    3.294713e+05
2    5.603352e+06
3    1.148324e+04
4    1.266668e+06
5    7.501556e+06
6    9.500199e+04
7    4.566914e+04
8    2.332056e+05
9    5.947277e+05
10   1.327782e+05
11   1.461058e+04
12   8.476088e+04
13   1.256888e+06
14   1.056287e+05
15   1.164235e+05
16   6.528548e+05
17   3.584550e+05

```

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```
18      9.655062e+04
19      1.305046e+06
20      8.990870e+04
21      1.691351e+05
22      1.269751e+06
Name: tCH4_2019, dtype: float64
```

```
results = stats.shapiro(data_2019)
print('stat=%.3f, p=%.3f' % (results.statistic, results.pvalue))
if results.pvalue > 0.05:
    print('Probably Gaussian')
else:
    print('Probably not Gaussian')
```

```
stat=0.567, p=0.000
Probably not Gaussian
```

## Results

The distribution is not gaussian so a non-parametric test must be completed. It is not necessary to perform this test on the 2020 data, but I will do so anyways for practice.

## 2020

```
## Selecting the Malaysia Data 2020
data_2020 = hypothesis_testing_df['tCH4_2020']
```

```
results = stats.shapiro(data_2020)
print('stat=%.3f, p=%.3f' % (results.statistic, results.pvalue))
if results.pvalue > 0.05:
    print('Probably Gaussian')
else:
    print('Probably not Gaussian')
```

```
stat=0.565, p=0.000
Probably not Gaussian
```

## Results

The 2020 data is not gaussian which verifies that we will need to perform a non parametric test

### 4.3.3 Independence of Samples.

We have to assume that the samples are independent of each other as we know they are dependent on hectares. Though the correlations are rather high this is due to the similarity of hectares per year. Thus the amount of ch4 is similar

### 4.3.4 Distribution Similarity

#### Mann-Whitney U Test

```
# Example of the Mann-Whitney U Test

stat, p = stats.mannwhitneyu(data_2019, data_2020)
print('stat=%.3f, p=%.3f' % (stat, p))
if p > 0.05:
    print('Probably the same distribution')
else:
    print('Probably different distributions')
```

```
stat=266.000, p=0.982
Probably the same distribution
```

### 4.3.5 Kruskal Wallis test

```
stat, p = stats.kruskal(data_2019, data_2020)
print('stat=%.3f, p=%.3f' % (stat, p))
if p > 0.05:
    print('Probably the same distribution')
else:
    print('Probably different distributions')
```

```
stat=0.001, p=0.974
Probably the same distribution
```

### 4.3.6 Friedman Test

Just for the sake of it I will compare data across all distributions

```
# Example of the Friedman Test
#data_2014 = hypothesis_testing_df['tCH4_2014']
data_2015 = hypothesis_testing_df['tCH4_2015']
data_2016 = hypothesis_testing_df['tCH4_2016']
data_2017 = hypothesis_testing_df['tCH4_2017']
data_2018 = hypothesis_testing_df['tCH4_2018']

stat, p = stats.friedmanchisquare(data_2015, data_2016, data_2017, data_2018, data_
    2019, data_2020)
print('stat=%.3f, p=%.3f' % (stat, p))
if p > 0.05:
    print('Probably the same distribution')
```

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```
else:  
    print('Probably different distributions')
```

```
stat=11.472, p=0.043  
Probably different distributions
```

### Results.

Some distributions differ from one another. Which those are have yet to be discovered. For the sake of this analysis I will not attempt to identify them.

The statment that the distributions of the 2019 and 2020 data do not differ cannot differ. That said we also cannot claim that the means are statistically equivalent as the data is not parametric.

# **Part III**

## **Data Exploration**





## EMISSIONS ESTIMATION DATA: A COMPARISON BETWEEN FAOSTAT AND UNIVERSITY OF MALYSIA ESTIMATES

### 5.1 Import

```
import pandas as pd
import matplotlib.pyplot as plt
import geopandas as gpd
import folium
import contextily as cx
from shapely.geometry import Point, LineString, Polygon
import numpy as np
from scipy.spatial import cKDTree
from geopy.distance import distance
import scipy.stats as stats
```

```
/Users/jnapolitano/venvs/finance/lib/python3.9/site-packages/geopandas/_compat.
↳py:111: UserWarning: The Shapely GEOS version (3.10.2-CAPI-1.16.0) is
↳incompatible with the GEOS version PyGEOS was compiled with (3.10.1-CAPI-1.16.0).
↳ Conversions between both will be slow.
warnings.warn(
```

#### 5.1.1 Dependencies

- Geopandas
- pandas
- openpyxl

### 5.2 Exploration Plan

#### 5.2.1 Data Imports

- /Users/jnapolitano/Projects/wattime-takehome/data/ch4\_2015-2021.xlsx
- /Users/jnapolitano/Projects/wattime-takehome/data/emissions\_csv\_fao\_emiss\_csv\_ch4\_fao\_2015\_2019\_tonnes.xlsx

## 5.2.2 Import Data Frames

Since jupyter caches the data to the notebook json I can import the dataframes that I will be using together.

If I were to build automated scripts to perform the analysis I would only load the data necessary to perform a process.

## 5.2.3 Experiment with Plots for each Set

I don't know exactly which plots I want to include in the final report.

I'll plot a few for each data set

## 5.2.4 Calculate differences between the datasets

- create a differences data frame
- write to file for use
- plot

## 5.3 University of Malaysia Emission Estimates

```
filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/ch4_2015-2021.xlsx"
malaysia_emissions_df = pd.read_excel(filepath)
```

### 5.3.1 Print Df Head

```
malaysia_emissions_df
```

	iso3_country	country_name	tCH4_2015	\
0	BGD	Bangladesh	2.344420e+06	
1	BRA	Brazil	3.410233e+05	
2	CHN	China	6.133647e+06	
3	ESP	Spain	1.141464e+04	
4	IDN	Indonesia	1.283649e+06	
5	IND	India	6.219887e+06	
6	IRN	Iran (Islamic Republic of)	8.774407e+04	
7	ITA	Italy	4.995968e+04	
8	JPN	Japan	2.305465e+05	
9	KHM	Cambodia	4.954698e+05	
10	KOR	Korea (the Republic of)	1.451878e+05	
11	LAO	Lao People's Democratic Republic (the)	1.661169e+04	
12	LKA	Sri Lanka	8.305626e+04	
13	MMR	Myanmar	1.132082e+06	
14	MYS	Malaysia	1.057399e+05	
15	NPL	Nepal	1.007479e+05	
16	PAK	Pakistan	4.852431e+05	
17	PHL	Philippines (the)	3.432021e+05	
18	PRK	Korea (the Democratic People's Republic of)	1.143217e+05	

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19	THA	Thailand	1.393798e+06
20	TWN	Taiwan (Province of China)	7.866956e+04
21	USA	United States of America (the)	1.611324e+05
22	VNM	Viet Nam	1.346013e+06
23	NaN	Total	2.270357e+07

	tCH4_2016	tCH4_2017	tCH4_2018	tCH4_2019	tCH4_2020 \
0	2.278158e+06	2.098958e+06	2.141231e+06	2.070985e+06	2.106781e+06
1	3.104189e+05	3.725173e+05	3.717030e+05	3.294713e+05	4.902874e+05
2	5.859531e+06	6.355071e+06	5.413962e+06	5.603352e+06	6.402353e+06
3	1.334803e+04	1.217299e+04	1.405410e+04	1.148324e+04	1.305461e+04
4	1.023129e+06	9.615327e+05	1.176982e+06	1.266668e+06	1.188195e+06
5	5.309413e+06	6.228451e+06	6.589798e+06	7.501556e+06	7.599764e+06
6	9.180121e+04	9.620217e+04	8.875744e+04	9.500199e+04	9.600254e+04
7	4.937785e+04	5.443679e+04	4.469902e+04	4.566914e+04	5.101547e+04
8	2.284133e+05	2.708935e+05	1.548252e+05	2.332056e+05	2.835167e+05
9	5.731698e+05	4.517045e+05	5.592610e+05	5.947277e+05	6.412802e+05
10	1.274597e+05	1.463222e+05	1.293543e+05	1.327782e+05	1.165467e+05
11	1.696441e+04	1.168063e+04	1.009675e+04	1.461058e+04	2.136270e+04
12	1.011743e+05	5.911841e+04	9.018914e+04	8.476088e+04	9.248238e+04
13	1.290806e+06	1.205169e+06	1.372447e+06	1.256888e+06	1.221904e+06
14	1.110049e+05	1.111291e+05	1.066525e+05	1.056287e+05	1.127141e+05
15	6.667161e+04	8.081300e+04	9.200752e+04	1.164235e+05	7.168401e+04
16	5.945922e+05	5.372641e+05	4.532297e+05	6.528548e+05	6.401201e+05
17	4.073554e+05	3.836830e+05	4.175210e+05	3.584550e+05	4.462836e+05
18	9.177653e+04	1.085457e+05	8.662578e+04	9.655062e+04	8.581038e+04
19	1.780993e+06	1.164699e+06	9.166575e+05	1.305046e+06	1.520788e+06
20	8.089149e+04	8.705634e+04	8.138151e+04	8.990870e+04	8.333327e+04
21	1.618576e+05	1.684799e+05	1.657254e+05	1.691351e+05	1.941455e+05
22	1.483777e+06	1.406437e+06	1.317455e+06	1.269751e+06	1.374450e+06
23	2.205208e+07	2.237234e+07	2.179462e+07	2.340491e+07	2.485387e+07

	tCH4_2021
0	1.983974e+06
1	4.544874e+05
2	6.068210e+06
3	8.531579e+03
4	1.009936e+06
5	6.567960e+06
6	9.053525e+04
7	5.089759e+04
8	1.574007e+05
9	5.644891e+05
10	1.013006e+05
11	1.475014e+04
12	8.466966e+04
13	1.289837e+06
14	1.069696e+05
15	4.811408e+04
16	4.849205e+05
17	4.383270e+05
18	7.735988e+04
19	8.528673e+05
20	6.619861e+04
21	1.634842e+05
22	1.502787e+06
23	2.218801e+07

### 5.3.2 Calculate Co2 Equivalency

```
malaysia_emissions_df['tCO2_2015'] = (malaysia_emissions_df['tCH4_2015'] * 25)
malaysia_emissions_df['tCO2_2016'] = (malaysia_emissions_df['tCH4_2016'] * 25)
malaysia_emissions_df['tCO2_2017'] = (malaysia_emissions_df['tCH4_2017'] * 25)
malaysia_emissions_df['tCO2_2018'] = (malaysia_emissions_df['tCH4_2018'] * 25)
malaysia_emissions_df['tCO2_2019'] = (malaysia_emissions_df['tCH4_2019'] * 25)
```

### 5.3.3 Calculate Means

```
malaysia_emissions_df.loc['mean'] = malaysia_emissions_df.loc[(malaysia_emissions_df[
    'country_name'] != "Total")].select_dtypes(np.number).mean()
malaysia_emissions_df.at['mean', 'country_name'] = 'mean'
malaysia_emissions_df
```

	iso3_country	country_name	tCH4_2015	\
0	BGD	Bangladesh	2.344420e+06	
1	BRA	Brazil	3.410233e+05	
2	CHN	China	6.133647e+06	
3	ESP	Spain	1.141464e+04	
4	IDN	Indonesia	1.283649e+06	
5	IND	India	6.219887e+06	
6	IRN	Iran (Islamic Republic of)	8.774407e+04	
7	ITA	Italy	4.995968e+04	
8	JPN	Japan	2.305465e+05	
9	KHM	Cambodia	4.954698e+05	
10	KOR	Korea (the Republic of)	1.451878e+05	
11	LAO	Lao People's Democratic Republic (the)	1.661169e+04	
12	LKA	Sri Lanka	8.305626e+04	
13	MMR	Myanmar	1.132082e+06	
14	MYS	Malaysia	1.057399e+05	
15	NPL	Nepal	1.007479e+05	
16	PAK	Pakistan	4.852431e+05	
17	PHL	Philippines (the)	3.432021e+05	
18	PRK	Korea (the Democratic People's Republic of)	1.143217e+05	
19	THA	Thailand	1.393798e+06	
20	TWN	Taiwan (Province of China)	7.866956e+04	
21	USA	United States of America (the)	1.611324e+05	
22	VNM	Viet Nam	1.346013e+06	
23	NaN	Total	2.270357e+07	
mean	NaN	mean	9.871116e+05	

	tCH4_2016	tCH4_2017	tCH4_2018	tCH4_2019	tCH4_2020	\
0	2.278158e+06	2.098958e+06	2.141231e+06	2.070985e+06	2.106781e+06	
1	3.104189e+05	3.725173e+05	3.717030e+05	3.294713e+05	4.902874e+05	
2	5.859531e+06	6.355071e+06	5.413962e+06	5.603352e+06	6.402353e+06	
3	1.334803e+04	1.217299e+04	1.405410e+04	1.148324e+04	1.305461e+04	
4	1.023129e+06	9.615327e+05	1.176982e+06	1.266668e+06	1.188195e+06	
5	5.309413e+06	6.228451e+06	6.589798e+06	7.501556e+06	7.599764e+06	
6	9.180121e+04	9.620217e+04	8.875744e+04	9.500199e+04	9.600254e+04	
7	4.937785e+04	5.443679e+04	4.469902e+04	4.566914e+04	5.101547e+04	
8	2.284133e+05	2.708935e+05	1.548252e+05	2.332056e+05	2.835167e+05	
9	5.731698e+05	4.517045e+05	5.592610e+05	5.947277e+05	6.412802e+05	
10	1.274597e+05	1.463222e+05	1.293543e+05	1.327782e+05	1.165467e+05	

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11	1.696441e+04	1.168063e+04	1.009675e+04	1.461058e+04	2.136270e+04
12	1.011743e+05	5.911841e+04	9.018914e+04	8.476088e+04	9.248238e+04
13	1.290806e+06	1.205169e+06	1.372447e+06	1.256888e+06	1.221904e+06
14	1.110049e+05	1.111291e+05	1.066525e+05	1.056287e+05	1.127141e+05
15	6.667161e+04	8.081300e+04	9.200752e+04	1.164235e+05	7.168401e+04
16	5.945922e+05	5.372641e+05	4.532297e+05	6.528548e+05	6.401201e+05
17	4.073554e+05	3.836830e+05	4.175210e+05	3.584550e+05	4.462836e+05
18	9.177653e+04	1.085457e+05	8.662578e+04	9.655062e+04	8.581038e+04
19	1.780993e+06	1.164699e+06	9.166575e+05	1.305046e+06	1.520788e+06
20	8.089149e+04	8.705634e+04	8.138151e+04	8.990870e+04	8.333327e+04
21	1.618576e+05	1.684799e+05	1.657254e+05	1.691351e+05	1.941455e+05
22	1.483777e+06	1.406437e+06	1.317455e+06	1.269751e+06	1.374450e+06
23	2.205208e+07	2.237234e+07	2.179462e+07	2.340491e+07	2.485387e+07
mean	9.587863e+05	9.727103e+05	9.475920e+05	1.017605e+06	1.080603e+06

	tCH4_2021	tCO2_2015	tCO2_2016	tCO2_2017	tCO2_2018	\
0	1.983974e+06	5.861049e+07	5.695395e+07	5.247394e+07	5.353076e+07	
1	4.544874e+05	8.525583e+06	7.760473e+06	9.312934e+06	9.292575e+06	
2	6.068210e+06	1.533412e+08	1.464883e+08	1.588768e+08	1.353491e+08	
3	8.531579e+03	2.853661e+05	3.337007e+05	3.043248e+05	3.513524e+05	
4	1.009936e+06	3.209122e+07	2.557824e+07	2.403832e+07	2.942454e+07	
5	6.567960e+06	1.554972e+08	1.327353e+08	1.557113e+08	1.647450e+08	
6	9.053525e+04	2.193602e+06	2.295030e+06	2.405054e+06	2.218936e+06	
7	5.089759e+04	1.248992e+06	1.234446e+06	1.360920e+06	1.117475e+06	
8	1.574007e+05	5.763662e+06	5.710333e+06	6.772337e+06	3.870631e+06	
9	5.644891e+05	1.238675e+07	1.432925e+07	1.129261e+07	1.398153e+07	
10	1.013006e+05	3.629695e+06	3.186493e+06	3.658056e+06	3.233858e+06	
11	1.475014e+04	4.152924e+05	4.241102e+05	2.920158e+05	2.524186e+05	
12	8.466966e+04	2.076407e+06	2.529358e+06	1.477960e+06	2.254728e+06	
13	1.289837e+06	2.830206e+07	3.227014e+07	3.012923e+07	3.431117e+07	
14	1.069696e+05	2.643498e+06	2.775123e+06	2.778227e+06	2.666313e+06	
15	4.811408e+04	2.518697e+06	1.666790e+06	2.020325e+06	2.300188e+06	
16	4.849205e+05	1.213108e+07	1.486480e+07	1.343160e+07	1.133074e+07	
17	4.383270e+05	8.580052e+06	1.018389e+07	9.592074e+06	1.043803e+07	
18	7.735988e+04	2.858041e+06	2.294413e+06	2.713641e+06	2.165645e+06	
19	8.528673e+05	3.484495e+07	4.452483e+07	2.911748e+07	2.291644e+07	
20	6.619861e+04	1.966739e+06	2.022287e+06	2.176408e+06	2.034538e+06	
21	1.634842e+05	4.028310e+06	4.046440e+06	4.211999e+06	4.143136e+06	
22	1.502787e+06	3.365033e+07	3.709441e+07	3.516092e+07	3.293637e+07	
23	2.218801e+07	5.675891e+08	5.513021e+08	5.593084e+08	5.448654e+08	
mean	9.646959e+05	2.467779e+07	2.396966e+07	2.431776e+07	2.368980e+07	

	tCO2_2019
0	5.177463e+07
1	8.236783e+06
2	1.400838e+08
3	2.870810e+05
4	3.166670e+07
5	1.875389e+08
6	2.375050e+06
7	1.141729e+06
8	5.830141e+06
9	1.486819e+07
10	3.319455e+06
11	3.652645e+05
12	2.119022e+06

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```

13      3.142221e+07
14      2.640717e+06
15      2.910588e+06
16      1.632137e+07
17      8.961374e+06
18      2.413765e+06
19      3.262615e+07
20      2.247717e+06
21      4.228377e+06
22      3.174377e+07
23      5.851228e+08
mean    2.544012e+07

```

### 5.3.4 Calculate Means and Totals Across Rows

```

mean_series = malaysia_emissions_df[['tCH4_2015', 'tCH4_2016', 'tCH4_2017', 'tCH4_2018',
↪ 'tCH4_2019']].select_dtypes(np.number).mean(axis=1)
total_series = malaysia_emissions_df[['tCH4_2015', 'tCH4_2016', 'tCH4_2017', 'tCH4_2018',
↪ 'tCH4_2019']].select_dtypes(np.number).sum(axis=1)
malaysia_emissions_df["Mean_CH4"] = mean_series
malaysia_emissions_df["Total_CH4"] = total_series

```

```

## the select np.number is unnecessary, but i'm including anyways as it doesnt really_
↪ hurt but for a small calculation penalty
mean_series = malaysia_emissions_df[['tCO2_2015', 'tCO2_2016', 'tCO2_2017', 'tCO2_2018',
↪ 'tCO2_2019']].select_dtypes(np.number).mean(axis=1)
total_series = malaysia_emissions_df[['tCO2_2015', 'tCO2_2016', 'tCO2_2017', 'tCO2_2018',
↪ 'tCO2_2019']].select_dtypes(np.number).sum(axis=1)
malaysia_emissions_df["Mean_CO2"] = mean_series
malaysia_emissions_df["Total_CO2"] = total_series

```

```
malaysia_emissions_df.reset_index(inplace=True, drop = True)
```

```
malaysia_emissions_df
```

	iso3_country	country_name	tCH4_2015 \
0	BGD	Bangladesh	2.344420e+06
1	BRA	Brazil	3.410233e+05
2	CHN	China	6.133647e+06
3	ESP	Spain	1.141464e+04
4	IDN	Indonesia	1.283649e+06
5	IND	India	6.219887e+06
6	IRN	Iran (Islamic Republic of)	8.774407e+04
7	ITA	Italy	4.995968e+04
8	JPN	Japan	2.305465e+05
9	KHM	Cambodia	4.954698e+05
10	KOR	Korea (the Republic of)	1.451878e+05
11	LAO	Lao People's Democratic Republic (the)	1.661169e+04
12	LKA	Sri Lanka	8.305626e+04
13	MMR	Myanmar	1.132082e+06
14	MYS	Malaysia	1.057399e+05

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15	NPL	Nepal	1.007479e+05
16	PAK	Pakistan	4.852431e+05
17	PHL	Philippines (the)	3.432021e+05
18	PRK	Korea (the Democratic People's Republic of)	1.143217e+05
19	THA	Thailand	1.393798e+06
20	TWN	Taiwan (Province of China)	7.866956e+04
21	USA	United States of America (the)	1.611324e+05
22	VNM	Viet Nam	1.346013e+06
23	NaN	Total	2.270357e+07
24	NaN	mean	9.871116e+05

	tCH4_2016	tCH4_2017	tCH4_2018	tCH4_2019	tCH4_2020 \
0	2.278158e+06	2.098958e+06	2.141231e+06	2.070985e+06	2.106781e+06
1	3.104189e+05	3.725173e+05	3.717030e+05	3.294713e+05	4.902874e+05
2	5.859531e+06	6.355071e+06	5.413962e+06	5.603352e+06	6.402353e+06
3	1.334803e+04	1.217299e+04	1.405410e+04	1.148324e+04	1.305461e+04
4	1.023129e+06	9.615327e+05	1.176982e+06	1.266668e+06	1.188195e+06
5	5.309413e+06	6.228451e+06	6.589798e+06	7.501556e+06	7.599764e+06
6	9.180121e+04	9.620217e+04	8.875744e+04	9.500199e+04	9.600254e+04
7	4.937785e+04	5.443679e+04	4.469902e+04	4.566914e+04	5.101547e+04
8	2.284133e+05	2.708935e+05	1.548252e+05	2.332056e+05	2.835167e+05
9	5.731698e+05	4.517045e+05	5.592610e+05	5.947277e+05	6.412802e+05
10	1.274597e+05	1.463222e+05	1.293543e+05	1.327782e+05	1.165467e+05
11	1.696441e+04	1.168063e+04	1.009675e+04	1.461058e+04	2.136270e+04
12	1.011743e+05	5.911841e+04	9.018914e+04	8.476088e+04	9.248238e+04
13	1.290806e+06	1.205169e+06	1.372447e+06	1.256888e+06	1.221904e+06
14	1.110049e+05	1.111291e+05	1.066525e+05	1.056287e+05	1.127141e+05
15	6.667161e+04	8.081300e+04	9.200752e+04	1.164235e+05	7.168401e+04
16	5.945922e+05	5.372641e+05	4.532297e+05	6.528548e+05	6.401201e+05
17	4.073554e+05	3.836830e+05	4.175210e+05	3.584550e+05	4.462836e+05
18	9.177653e+04	1.085457e+05	8.662578e+04	9.655062e+04	8.581038e+04
19	1.780993e+06	1.164699e+06	9.166575e+05	1.305046e+06	1.520788e+06
20	8.089149e+04	8.705634e+04	8.138151e+04	8.990870e+04	8.333327e+04
21	1.618576e+05	1.684799e+05	1.657254e+05	1.691351e+05	1.941455e+05
22	1.483777e+06	1.406437e+06	1.317455e+06	1.269751e+06	1.374450e+06
23	2.205208e+07	2.237234e+07	2.179462e+07	2.340491e+07	2.485387e+07
24	9.587863e+05	9.727103e+05	9.475920e+05	1.017605e+06	1.080603e+06

	tCH4_2021	tCO2_2015	tCO2_2016	tCO2_2017	tCO2_2018 \
0	1.983974e+06	5.861049e+07	5.695395e+07	5.247394e+07	5.353076e+07
1	4.544874e+05	8.525583e+06	7.760473e+06	9.312934e+06	9.292575e+06
2	6.068210e+06	1.533412e+08	1.464883e+08	1.588768e+08	1.353491e+08
3	8.531579e+03	2.853661e+05	3.337007e+05	3.043248e+05	3.513524e+05
4	1.009936e+06	3.209122e+07	2.557824e+07	2.403832e+07	2.942454e+07
5	6.567960e+06	1.554972e+08	1.327353e+08	1.557113e+08	1.647450e+08
6	9.053525e+04	2.193602e+06	2.295030e+06	2.405054e+06	2.218936e+06
7	5.089759e+04	1.248992e+06	1.234446e+06	1.360920e+06	1.117475e+06
8	1.574007e+05	5.763662e+06	5.710333e+06	6.772337e+06	3.870631e+06
9	5.644891e+05	1.238675e+07	1.432925e+07	1.129261e+07	1.398153e+07
10	1.013006e+05	3.629695e+06	3.186493e+06	3.658056e+06	3.233858e+06
11	1.475014e+04	4.152924e+05	4.241102e+05	2.920158e+05	2.524186e+05
12	8.466966e+04	2.076407e+06	2.529358e+06	1.477960e+06	2.254728e+06
13	1.289837e+06	2.830206e+07	3.227014e+07	3.012923e+07	3.431117e+07
14	1.069696e+05	2.643498e+06	2.775123e+06	2.778227e+06	2.666313e+06
15	4.811408e+04	2.518697e+06	1.666790e+06	2.020325e+06	2.300188e+06
16	4.849205e+05	1.213108e+07	1.486480e+07	1.343160e+07	1.133074e+07

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17	4.383270e+05	8.580052e+06	1.018389e+07	9.592074e+06	1.043803e+07
18	7.735988e+04	2.858041e+06	2.294413e+06	2.713641e+06	2.165645e+06
19	8.528673e+05	3.484495e+07	4.452483e+07	2.911748e+07	2.291644e+07
20	6.619861e+04	1.966739e+06	2.022287e+06	2.176408e+06	2.034538e+06
21	1.634842e+05	4.028310e+06	4.046440e+06	4.211999e+06	4.143136e+06
22	1.502787e+06	3.365033e+07	3.709441e+07	3.516092e+07	3.293637e+07
23	2.218801e+07	5.675891e+08	5.513021e+08	5.593084e+08	5.448654e+08
24	9.646959e+05	2.467779e+07	2.396966e+07	2.431776e+07	2.368980e+07

	tCO2_2019	Mean_CH4	Total_CH4	Mean_CO2	Total_CO2
0	5.177463e+07	2.186750e+06	1.093375e+07	5.466875e+07	2.733438e+08
1	8.236783e+06	3.450268e+05	1.725134e+06	8.625670e+06	4.312835e+07
2	1.400838e+08	5.873113e+06	2.936556e+07	1.468278e+08	7.341391e+08
3	2.870810e+05	1.249460e+04	6.247300e+04	3.123650e+05	1.561825e+06
4	3.166670e+07	1.142392e+06	5.711960e+06	2.855980e+07	1.427990e+08
5	1.875389e+08	6.369821e+06	3.184910e+07	1.592455e+08	7.962276e+08
6	2.375050e+06	9.190138e+04	4.595069e+05	2.297534e+06	1.148767e+07
7	1.141729e+06	4.882850e+04	2.441425e+05	1.220712e+06	6.103562e+06
8	5.830141e+06	2.235768e+05	1.117884e+06	5.589421e+06	2.794710e+07
9	1.486819e+07	5.348666e+05	2.674333e+06	1.337166e+07	6.685832e+07
10	3.319455e+06	1.362205e+05	6.811023e+05	3.405512e+06	1.702756e+07
11	3.652645e+05	1.399281e+04	6.996406e+04	3.498203e+05	1.749102e+06
12	2.119022e+06	8.365981e+04	4.182990e+05	2.091495e+06	1.045748e+07
13	3.142221e+07	1.251478e+06	6.257392e+06	3.128696e+07	1.564348e+08
14	2.640717e+06	1.080310e+05	5.401551e+05	2.700775e+06	1.350388e+07
15	2.910588e+06	9.133271e+04	4.566635e+05	2.283318e+06	1.141659e+07
16	1.632137e+07	5.446368e+05	2.723184e+06	1.361592e+07	6.807960e+07
17	8.961374e+06	3.820433e+05	1.910216e+06	9.551082e+06	4.775541e+07
18	2.413765e+06	9.956405e+04	4.978202e+05	2.489101e+06	1.244551e+07
19	3.262615e+07	1.312239e+06	6.561194e+06	3.280597e+07	1.640298e+08
20	2.247717e+06	8.358152e+04	4.179076e+05	2.089538e+06	1.044769e+07
21	4.228377e+06	1.652661e+05	8.263305e+05	4.131652e+06	2.065826e+07
22	3.174377e+07	1.364686e+06	6.823432e+06	3.411716e+07	1.705858e+08
23	5.851228e+08	2.246550e+07	1.123275e+08	5.616376e+08	2.808188e+09
24	2.544012e+07	9.767610e+05	4.883805e+06	2.441902e+07	1.220951e+08

### 5.3.5 Write Data to File

```
outfile = "/Users/jnapolitano/Projects/watttime-takehome/data/TRACE_DATA.csv"

malaysia_emissions_df.to_csv(outfile)
```

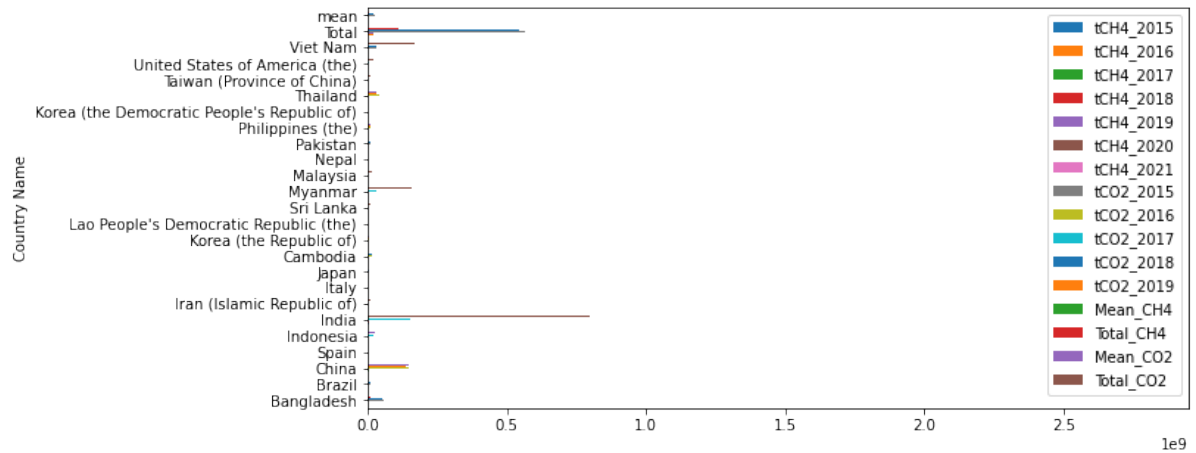
## 5.4 University of Malaysia Plots

### 5.4.1 University of Malaysia Bar Plot

```
malaysia_emissions_df.plot(kind = "barh", x = 'country_name', xlabel = "Country Name",
↪ ylabel = "CH4 Tonnes", figsize = (10,5))
```



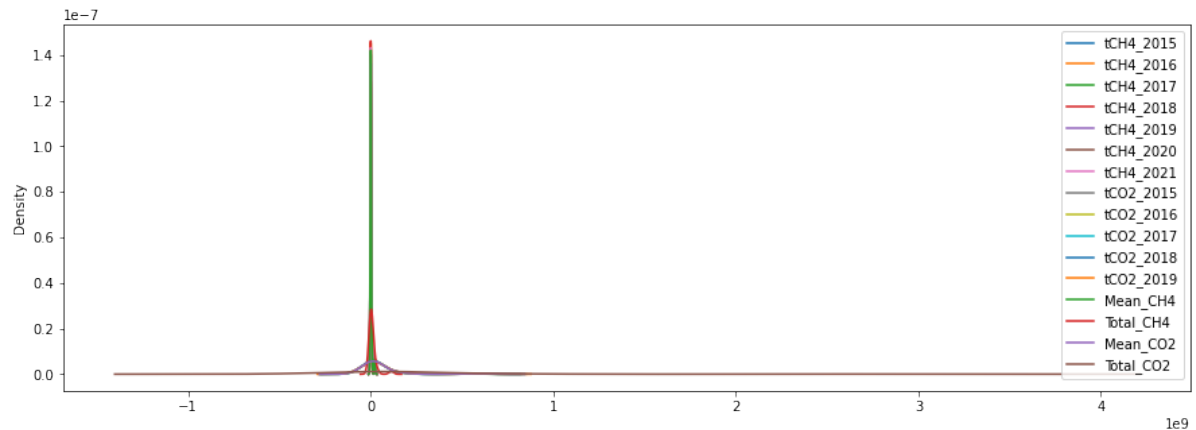
```
<AxesSubplot:ylabel='Country Name'>
```



## 5.4.2 University of Malaysia Density Plot

```
malaysia_emissions_df.plot(rot = 0, kind = "density", figsize = (15,5))
```

```
<AxesSubplot:ylabel='Density'>
```



I did not exclude totals or mean from the dataframe, but as we can see the second hump in the density graph shows the distribution of totals annually. Interestingly the 2020 data is shifted further to the right than other years. This actually questions the validity of the study promoted by the University of Malaysia

## 5.5 FAOSTAT Data

```
filepath = "/Users/jnapolitano/Projects/wattime-takehome/data/emissions_csv_fao_emiss_
↪csv_ch4_fao_2015_2019_tonnes.xlsx"

faostat_emissions_df = pd.read_excel(filepath)
```

### 5.5.1 Print FAOSTAT Data

```
## I didn't write the index to the csv file in the previous step. IF time permits go
↪back and fix this error
faostat_emissions_df
```

	code	country \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India
6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand
20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2015	2016	2017 \
0	Bangladesh	1131293.4	1093480.4	1154531.0
1	Brazil	138910.3	126278.2	130322.9
2	China, mainland	5406593.9	5399920.0	5400129.0
3	Spain	55082.2	55073.1	54232.4
4	Indonesia	2407953.5	2387656.4	2425290.6
5	India	4580248.4	4559136.4	4620790.8
6	Iran (Islamic Republic of)	116486.7	131008.5	87233.6
7	Italy	114574.8	118003.0	118003.0
8	Japan	330353.1	326403.0	323700.3
9	Cambodia	436826.0	459003.1	473745.3
10	Republic of Korea	167862.2	163534.1	158489.7
11	Lao People's Democratic Republic	94826.8	95630.0	93940.7
12	Sri Lanka	132640.0	121756.3	84456.3

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13	Myanmar	1059409.6	1052287.7	1087029.5
14	Malaysia	121942.6	123232.8	122656.3
15	Nepal	149262.2	142723.7	162574.6
16	Pakistan	383529.3	381361.8	406083.3
17	Philippines	1557810.6	1524292.5	1609862.5
18	Democratic People's Republic of Korea	82823.3	83442.3	84596.2
19	Thailand	1554254.0	1703327.7	1714465.6
20	China, Taiwan Province of	45838.7	49838.3	49991.2
21	United States of America	364728.0	438662.0	336255.5
22	Viet Nam	1381744.4	1365173.8	1360551.6
23	Total	23829994.0	23917225.1	24075931.9

	2018	2019	2020	2021
0	1144591.0	1144745.4	NaN	NaN
1	121615.2	111084.8	NaN	NaN
2	5302173.1	5214454.7	NaN	NaN
3	52925.0	52098.5	NaN	NaN
4	2405613.8	2257604.3	NaN	NaN
5	4661154.9	4621416.8	NaN	NaN
6	93936.6	96103.4	NaN	NaN
7	109463.8	110895.1	NaN	NaN
8	322245.0	320581.8	NaN	NaN
9	479362.7	468378.9	NaN	NaN
10	154911.3	153260.9	NaN	NaN
11	83333.6	77005.5	NaN	NaN
12	111049.0	102156.3	NaN	NaN
13	1118850.0	1083100.3	NaN	NaN
14	125238.5	122453.8	NaN	NaN
15	153890.8	156215.4	NaN	NaN
16	393404.2	424755.1	NaN	NaN
17	1606047.8	1556225.8	NaN	NaN
18	83943.3	82937.0	NaN	NaN
19	1702989.1	1553835.5	NaN	NaN
20	49414.1	49152.0	NaN	NaN
21	412177.5	350136.5	NaN	NaN
22	1336231.2	1318431.1	NaN	NaN
23	24042561.5	23446028.9	NaN	NaN

## 5.5.2 Change code to iso3\_country

```
faostat_emissions_df.rename(columns={"code": "iso3_country"}, inplace =True)
faostat_emissions_df.rename(columns={"country": "country_name"}, inplace =True)
# The column title is not a string. It is understood as an int or a datetime.
#faostat_emissions_df['2015']
faostat_emissions_df
```

	iso3_country	country_name \
0	BGD	Bangladesh
1	BRA	Brazil
2	CHN	China
3	ESP	Spain
4	IDN	Indonesia
5	IND	India

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6	IRN	Iran (Islamic Republic of)
7	ITA	Italy
8	JPN	Japan
9	KHM	Cambodia
10	KOR	Korea (the Republic of)
11	LAO	Lao People's Democratic Republic (the)
12	LKA	Sri Lanka
13	MMR	Myanmar
14	MYS	Malaysia
15	NPL	Nepal
16	PAK	Pakistan
17	PHL	Philippines (the)
18	PRK	Korea (the Democratic People's Republic of)
19	THA	Thailand
20	TWN	Taiwan (Province of China)
21	USA	United States of America (the)
22	VNM	Viet Nam
23	NaN	NaN

	country_fao	2015	2016	2017	\
0	Bangladesh	1131293.4	1093480.4	1154531.0	
1	Brazil	138910.3	126278.2	130322.9	
2	China, mainland	5406593.9	5399920.0	5400129.0	
3	Spain	55082.2	55073.1	54232.4	
4	Indonesia	2407953.5	2387656.4	2425290.6	
5	India	4580248.4	4559136.4	4620790.8	
6	Iran (Islamic Republic of)	116486.7	131008.5	87233.6	
7	Italy	114574.8	118003.0	118003.0	
8	Japan	330353.1	326403.0	323700.3	
9	Cambodia	436826.0	459003.1	473745.3	
10	Republic of Korea	167862.2	163534.1	158489.7	
11	Lao People's Democratic Republic	94826.8	95630.0	93940.7	
12	Sri Lanka	132640.0	121756.3	84456.3	
13	Myanmar	1059409.6	1052287.7	1087029.5	
14	Malaysia	121942.6	123232.8	122656.3	
15	Nepal	149262.2	142723.7	162574.6	
16	Pakistan	383529.3	381361.8	406083.3	
17	Philippines	1557810.6	1524292.5	1609862.5	
18	Democratic People's Republic of Korea	82823.3	83442.3	84596.2	
19	Thailand	1554254.0	1703327.7	1714465.6	
20	China, Taiwan Province of	45838.7	49838.3	49991.2	
21	United States of America	364728.0	438662.0	336255.5	
22	Viet Nam	1381744.4	1365173.8	1360551.6	
23	Total	23829994.0	23917225.1	24075931.9	

	2018	2019	2020	2021
0	1144591.0	1144745.4	NaN	NaN
1	121615.2	111084.8	NaN	NaN
2	5302173.1	5214454.7	NaN	NaN
3	52925.0	52098.5	NaN	NaN
4	2405613.8	2257604.3	NaN	NaN
5	4661154.9	4621416.8	NaN	NaN
6	93936.6	96103.4	NaN	NaN
7	109463.8	110895.1	NaN	NaN
8	322245.0	320581.8	NaN	NaN
9	479362.7	468378.9	NaN	NaN

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10	154911.3	153260.9	NaN	NaN
11	83333.6	77005.5	NaN	NaN
12	111049.0	102156.3	NaN	NaN
13	1118850.0	1083100.3	NaN	NaN
14	125238.5	122453.8	NaN	NaN
15	153890.8	156215.4	NaN	NaN
16	393404.2	424755.1	NaN	NaN
17	1606047.8	1556225.8	NaN	NaN
18	83943.3	82937.0	NaN	NaN
19	1702989.1	1553835.5	NaN	NaN
20	49414.1	49152.0	NaN	NaN
21	412177.5	350136.5	NaN	NaN
22	1336231.2	1318431.1	NaN	NaN
23	24042561.5	23446028.9	NaN	NaN

### 5.5.3 Set country\_name total to total

```
faostat_emissions_df.at[23, 'country_name'] = 'Total'
```

### 5.5.4 Drop Fao Country Code

```
faostat_emissions_df.drop(labels = ['country_fao'], axis=1, inplace=True)
faostat_emissions_df
```

	iso3_country	country_name	2015	\
0	BGD	Bangladesh	1131293.4	
1	BRA	Brazil	138910.3	
2	CHN	China	5406593.9	
3	ESP	Spain	55082.2	
4	IDN	Indonesia	2407953.5	
5	IND	India	4580248.4	
6	IRN	Iran (Islamic Republic of)	116486.7	
7	ITA	Italy	114574.8	
8	JPN	Japan	330353.1	
9	KHM	Cambodia	436826.0	
10	KOR	Korea (the Republic of)	167862.2	
11	LAO	Lao People's Democratic Republic (the)	94826.8	
12	LKA	Sri Lanka	132640.0	
13	MMR	Myanmar	1059409.6	
14	MYS	Malaysia	121942.6	
15	NPL	Nepal	149262.2	
16	PAK	Pakistan	383529.3	
17	PHL	Philippines (the)	1557810.6	
18	PRK	Korea (the Democratic People's Republic of)	82823.3	
19	THA	Thailand	1554254.0	
20	TWN	Taiwan (Province of China)	45838.7	
21	USA	United States of America (the)	364728.0	
22	VNM	Viet Nam	1381744.4	
23	NaN	Total	23829994.0	

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	2016	2017	2018	2019	2020	2021
0	1093480.4	1154531.0	1144591.0	1144745.4	NaN	NaN
1	126278.2	130322.9	121615.2	111084.8	NaN	NaN
2	5399920.0	5400129.0	5302173.1	5214454.7	NaN	NaN
3	55073.1	54232.4	52925.0	52098.5	NaN	NaN
4	2387656.4	2425290.6	2405613.8	2257604.3	NaN	NaN
5	4559136.4	4620790.8	4661154.9	4621416.8	NaN	NaN
6	131008.5	87233.6	93936.6	96103.4	NaN	NaN
7	118003.0	118003.0	109463.8	110895.1	NaN	NaN
8	326403.0	323700.3	322245.0	320581.8	NaN	NaN
9	459003.1	473745.3	479362.7	468378.9	NaN	NaN
10	163534.1	158489.7	154911.3	153260.9	NaN	NaN
11	95630.0	93940.7	83333.6	77005.5	NaN	NaN
12	121756.3	84456.3	111049.0	102156.3	NaN	NaN
13	1052287.7	1087029.5	1118850.0	1083100.3	NaN	NaN
14	123232.8	122656.3	125238.5	122453.8	NaN	NaN
15	142723.7	162574.6	153890.8	156215.4	NaN	NaN
16	381361.8	406083.3	393404.2	424755.1	NaN	NaN
17	1524292.5	1609862.5	1606047.8	1556225.8	NaN	NaN
18	83442.3	84596.2	83943.3	82937.0	NaN	NaN
19	1703327.7	1714465.6	1702989.1	1553835.5	NaN	NaN
20	49838.3	49991.2	49414.1	49152.0	NaN	NaN
21	438662.0	336255.5	412177.5	350136.5	NaN	NaN
22	1365173.8	1360551.6	1336231.2	1318431.1	NaN	NaN
23	23917225.1	24075931.9	24042561.5	23446028.9	NaN	NaN

## 5.5.5 Calculate Co2 Equivalency

```

faostat_emissions_df['tCO2_2015'] = faostat_emissions_df[2015] * 25
faostat_emissions_df['tCO2_2016'] = faostat_emissions_df[2016] * 25
faostat_emissions_df['tCO2_2017'] = faostat_emissions_df[2017] * 25
faostat_emissions_df['tCO2_2018'] = faostat_emissions_df[2018] * 25
faostat_emissions_df['tCO2_2019'] = faostat_emissions_df[2019] * 25

```

## 5.5.6 Calculate Means

```

faostat_emissions_df.loc['mean'] = faostat_emissions_df.loc[(faostat_emissions_df[
    'country_name'] != "Total")].select_dtypes(np.number).mean()
faostat_emissions_df.at['mean', 'country_name'] = 'mean'
faostat_emissions_df.reset_index(inplace=True, drop=True)
#faostat_emissions_df.at['mean', 'country_fao'] = 'mean'

```

### 5.5.7 Calculate Means and Totals Across Rows

```
mean_series = faostat_emissions_df[[2015,2016,2017,2018,2019]].select_dtypes(np.
↳number).mean(axis=1)
total_series = faostat_emissions_df[[2015,2016,2017,2018,2019]].select_dtypes(np.
↳number).sum(axis=1)
faostat_emissions_df["Mean_CH4"] = mean_series
faostat_emissions_df['Total_CH4'] = total_series
```

```
## the select np.number is unnecessary, but i'm including anyways as it doesnt really
↳hurt but for a small calculation penalty
mean_series = faostat_emissions_df[['tCO2_2015','tCO2_2016','tCO2_2017','tCO2_2018',
↳'tCO2_2019']].select_dtypes(np.number).mean(axis=1)
total_series = faostat_emissions_df[['tCO2_2015','tCO2_2016','tCO2_2017','tCO2_2018',
↳'tCO2_2019']].select_dtypes(np.number).sum(axis=1)
faostat_emissions_df["Mean_CO2"] = mean_series
faostat_emissions_df['Total_CO2'] = total_series
```

```
faostat_emissions_df.reset_index(inplace=True, drop=True)
```

```
faostat_emissions_df
```

	iso3_country	country_name	2015	\
0	BGD	Bangladesh	1131293.4	
1	BRA	Brazil	138910.3	
2	CHN	China	5406593.9	
3	ESP	Spain	55082.2	
4	IDN	Indonesia	2407953.5	
5	IND	India	4580248.4	
6	IRN	Iran (Islamic Republic of)	116486.7	
7	ITA	Italy	114574.8	
8	JPN	Japan	330353.1	
9	KHM	Cambodia	436826.0	
10	KOR	Korea (the Republic of)	167862.2	
11	LAO	Lao People's Democratic Republic (the)	94826.8	
12	LKA	Sri Lanka	132640.0	
13	MMR	Myanmar	1059409.6	
14	MYS	Malaysia	121942.6	
15	NPL	Nepal	149262.2	
16	PAK	Pakistan	383529.3	
17	PHL	Philippines (the)	1557810.6	
18	PRK	Korea (the Democratic People's Republic of)	82823.3	
19	THA	Thailand	1554254.0	
20	TWN	Taiwan (Province of China)	45838.7	
21	USA	United States of America (the)	364728.0	
22	VNM	Viet Nam	1381744.4	
23	NaN	Total	23829994.0	
24	NaN	mean	948478.0	

	2016	2017	2018	2019	2020	2021	\
0	1.093480e+06	1.154531e+06	1.144591e+06	1.144745e+06	NaN	NaN	
1	1.262782e+05	1.303229e+05	1.216152e+05	1.110848e+05	NaN	NaN	
2	5.399920e+06	5.400129e+06	5.302173e+06	5.214455e+06	NaN	NaN	
3	5.507310e+04	5.423240e+04	5.292500e+04	5.209850e+04	NaN	NaN	

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4	2.387656e+06	2.425291e+06	2.405614e+06	2.257604e+06	NaN	NaN
5	4.559136e+06	4.620791e+06	4.661155e+06	4.621417e+06	NaN	NaN
6	1.310085e+05	8.723360e+04	9.393660e+04	9.610340e+04	NaN	NaN
7	1.180030e+05	1.180030e+05	1.094638e+05	1.108951e+05	NaN	NaN
8	3.264030e+05	3.237003e+05	3.222450e+05	3.205818e+05	NaN	NaN
9	4.590031e+05	4.737453e+05	4.793627e+05	4.683789e+05	NaN	NaN
10	1.635341e+05	1.584897e+05	1.549113e+05	1.532609e+05	NaN	NaN
11	9.563000e+04	9.394070e+04	8.333360e+04	7.700550e+04	NaN	NaN
12	1.217563e+05	8.445630e+04	1.110490e+05	1.021563e+05	NaN	NaN
13	1.052288e+06	1.087030e+06	1.118850e+06	1.083100e+06	NaN	NaN
14	1.232328e+05	1.226563e+05	1.252385e+05	1.224538e+05	NaN	NaN
15	1.427237e+05	1.625746e+05	1.538908e+05	1.562154e+05	NaN	NaN
16	3.813618e+05	4.060833e+05	3.934042e+05	4.247551e+05	NaN	NaN
17	1.524292e+06	1.609862e+06	1.606048e+06	1.556226e+06	NaN	NaN
18	8.344230e+04	8.459620e+04	8.394330e+04	8.293700e+04	NaN	NaN
19	1.703328e+06	1.714466e+06	1.702989e+06	1.553836e+06	NaN	NaN
20	4.983830e+04	4.999120e+04	4.941410e+04	4.915200e+04	NaN	NaN
21	4.386620e+05	3.362555e+05	4.121775e+05	3.501365e+05	NaN	NaN
22	1.365174e+06	1.360552e+06	1.336231e+06	1.318431e+06	NaN	NaN
23	2.391723e+07	2.407593e+07	2.404256e+07	2.344603e+07	NaN	NaN
24	9.522272e+05	9.590840e+05	9.575896e+05	9.316100e+05	NaN	NaN

	tCO2_2015	tCO2_2016	tCO2_2017	tCO2_2018	tCO2_2019	\
0	28282335.0	2.733701e+07	2.886328e+07	2.861478e+07	2.861863e+07	
1	3472757.5	3.156955e+06	3.258072e+06	3.040380e+06	2.777120e+06	
2	135164847.5	1.349980e+08	1.350032e+08	1.325543e+08	1.303614e+08	
3	1377055.0	1.376828e+06	1.355810e+06	1.323125e+06	1.302462e+06	
4	60198837.5	5.969141e+07	6.063226e+07	6.014035e+07	5.644011e+07	
5	114506210.0	1.139784e+08	1.155198e+08	1.165289e+08	1.155354e+08	
6	2912167.5	3.275212e+06	2.180840e+06	2.348415e+06	2.402585e+06	
7	2864370.0	2.950075e+06	2.950075e+06	2.736595e+06	2.772378e+06	
8	8258827.5	8.160075e+06	8.092508e+06	8.056125e+06	8.014545e+06	
9	10920650.0	1.147508e+07	1.184363e+07	1.198407e+07	1.170947e+07	
10	4196555.0	4.088352e+06	3.962243e+06	3.872783e+06	3.831522e+06	
11	2370670.0	2.390750e+06	2.348518e+06	2.083340e+06	1.925138e+06	
12	3316000.0	3.043908e+06	2.111408e+06	2.776225e+06	2.553908e+06	
13	26485240.0	2.630719e+07	2.717574e+07	2.797125e+07	2.707751e+07	
14	3048565.0	3.080820e+06	3.066408e+06	3.130962e+06	3.061345e+06	
15	3731555.0	3.568093e+06	4.064365e+06	3.847270e+06	3.905385e+06	
16	9588232.5	9.534045e+06	1.015208e+07	9.835105e+06	1.061888e+07	
17	38945265.0	3.810731e+07	4.024656e+07	4.015120e+07	3.890564e+07	
18	2070582.5	2.086058e+06	2.114905e+06	2.098582e+06	2.073425e+06	
19	38856350.0	4.258319e+07	4.286164e+07	4.257473e+07	3.884589e+07	
20	1145967.5	1.245958e+06	1.249780e+06	1.235352e+06	1.228800e+06	
21	9118200.0	1.096655e+07	8.406388e+06	1.030444e+07	8.753412e+06	
22	34543610.0	3.412934e+07	3.401379e+07	3.340578e+07	3.296078e+07	
23	595749850.0	5.979306e+08	6.018983e+08	6.010640e+08	5.861507e+08	
24	23711950.0	2.380568e+07	2.397710e+07	2.393974e+07	2.329025e+07	

	Mean_CH4	Total_CH4	Mean_CO2	Total_CO2
0	1.133728e+06	5.668641e+06	2.834321e+07	1.417160e+08
1	1.256423e+05	6.282114e+05	3.141057e+06	1.570528e+07
2	5.344654e+06	2.672327e+07	1.336164e+08	6.680818e+08
3	5.388224e+04	2.694112e+05	1.347056e+06	6.735280e+06
4	2.376824e+06	1.188412e+07	5.942059e+07	2.971030e+08
5	4.608549e+06	2.304275e+07	1.152137e+08	5.760687e+08

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6	1.049538e+05	5.247688e+05	2.623844e+06	1.311922e+07
7	1.141879e+05	5.709397e+05	2.854698e+06	1.427349e+07
8	3.246566e+05	1.623283e+06	8.116416e+06	4.058208e+07
9	4.634632e+05	2.317316e+06	1.158658e+07	5.793290e+07
10	1.596116e+05	7.980582e+05	3.990291e+06	1.995146e+07
11	8.894732e+04	4.447366e+05	2.223683e+06	1.111842e+07
12	1.104116e+05	5.520579e+05	2.760290e+06	1.380145e+07
13	1.080135e+06	5.400677e+06	2.700339e+07	1.350169e+08
14	1.231048e+05	6.155240e+05	3.077620e+06	1.538810e+07
15	1.529333e+05	7.646667e+05	3.823334e+06	1.911667e+07
16	3.978267e+05	1.989134e+06	9.945668e+06	4.972834e+07
17	1.570848e+06	7.854239e+06	3.927120e+07	1.963560e+08
18	8.354842e+04	4.177421e+05	2.088710e+06	1.044355e+07
19	1.645774e+06	8.228872e+06	4.114436e+07	2.057218e+08
20	4.884686e+04	2.442343e+05	1.221172e+06	6.105858e+06
21	3.803919e+05	1.901960e+06	9.509798e+06	4.754899e+07
22	1.352426e+06	6.762132e+06	3.381066e+07	1.690533e+08
23	2.386235e+07	1.193117e+08	5.965587e+08	2.982794e+09
24	9.497978e+05	4.748989e+06	2.374494e+07	1.187247e+08

## 5.5.8 FAOSTAT Data to File

```
outfile = "/Users/jnapolitano/Projects/watttime-takehome/data/FAOSTAT_DATA.csv"

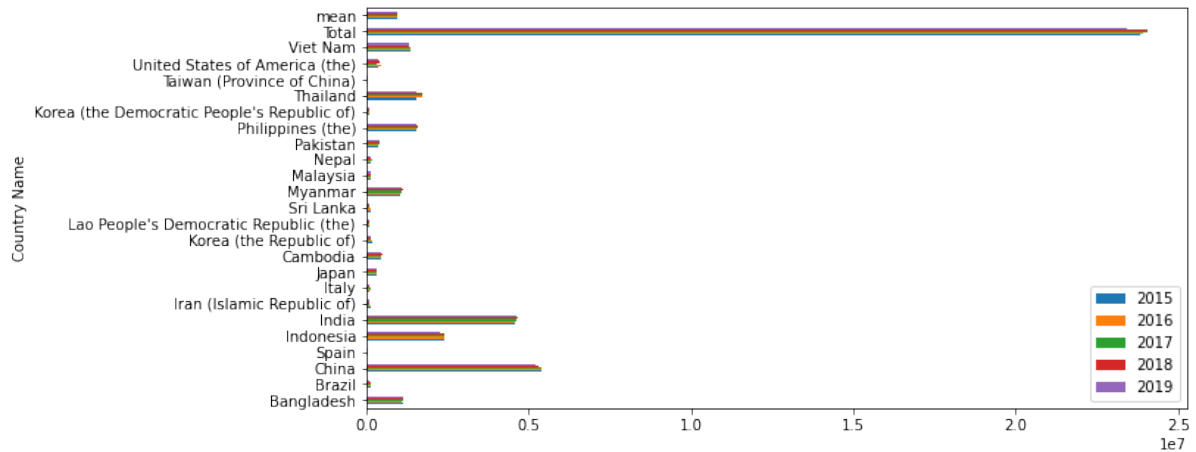
faostat_emissions_df.to_csv(outfile)
```

## 5.6 FaoSTAT PLOTS

### 5.6.1 FAOSTAT Hectare Estimates Bar Plot

```
faostat_emissions_df.plot(kind = "barh", x = 'country_name', y = [2015, 2016, 2017, ↵
↵2018, 2019], xlabel = "Country Name", ylabel = "Tonnes CH4", figsize = (10,5))
```

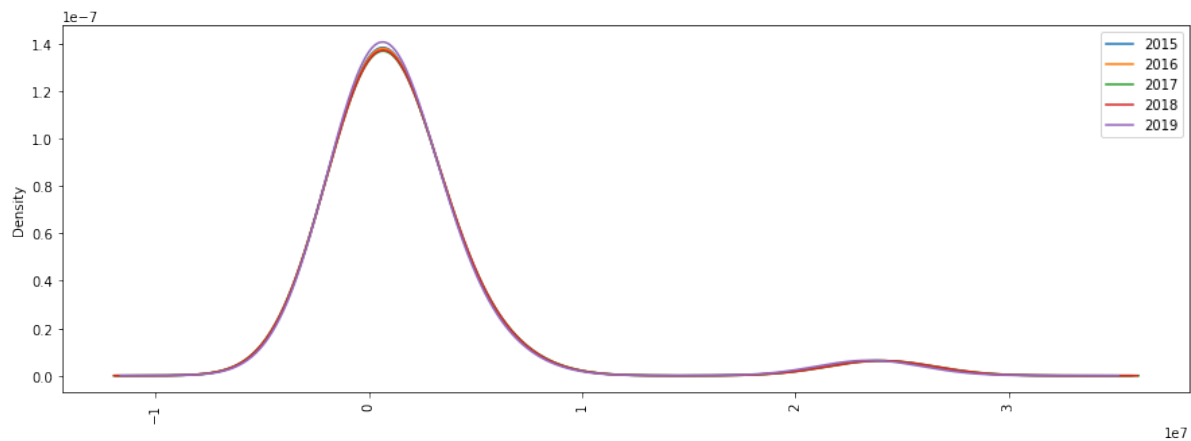
```
<AxesSubplot:ylabel='Country Name'>
```



### 5.6.2 FAOSTAT Density Plot

```
faostat_emissions_df.plot(rot = 90, kind = "density", y = [2015, 2016, 2017, 2018, 2019], figsize = (15,5))
```

```
<AxesSubplot:ylabel='Density'>
```



The density plot is fairly consistent. There is nearly no variation between nations and in total. The 2020 data may show otherwise as the Malaysian data shows.

## 5.7 Join Df's by ISO3 Country

### 5.7.1 Drop totals and means from the original df.

Because I am joining on iso3 country code if the totals and means are located at different indexes we may experience merge and calculation errors

```
faostat_emissions_df = faostat_emissions_df[(faostat_emissions_df["country_name"] != "Total") & (faostat_emissions_df['country_name'] != 'mean')].copy()
```

```
malaysia_emissions_df = malaysia_emissions_df[(malaysia_emissions_df["country_name"] !=
↪ "Total") & (malaysia_emissions_df['country_name'] != 'mean')].copy()
```

```
faostat_emissions_df
```

	iso3_country	country_name	2015	\
0	BGD	Bangladesh	1131293.4	
1	BRA	Brazil	138910.3	
2	CHN	China	5406593.9	
3	ESP	Spain	55082.2	
4	IDN	Indonesia	2407953.5	
5	IND	India	4580248.4	
6	IRN	Iran (Islamic Republic of)	116486.7	
7	ITA	Italy	114574.8	
8	JPN	Japan	330353.1	
9	KHM	Cambodia	436826.0	
10	KOR	Korea (the Republic of)	167862.2	
11	LAO	Lao People's Democratic Republic (the)	94826.8	
12	LKA	Sri Lanka	132640.0	
13	MMR	Myanmar	1059409.6	
14	MYS	Malaysia	121942.6	
15	NPL	Nepal	149262.2	
16	PAK	Pakistan	383529.3	
17	PHL	Philippines (the)	1557810.6	
18	PRK	Korea (the Democratic People's Republic of)	82823.3	
19	THA	Thailand	1554254.0	
20	TWN	Taiwan (Province of China)	45838.7	
21	USA	United States of America (the)	364728.0	
22	VNM	Viet Nam	1381744.4	

	2016	2017	2018	2019	2020	2021	tCO2_2015	\
0	1093480.4	1154531.0	1144591.0	1144745.4	NaN	NaN	28282335.0	
1	126278.2	130322.9	121615.2	111084.8	NaN	NaN	3472757.5	
2	5399920.0	5400129.0	5302173.1	5214454.7	NaN	NaN	135164847.5	
3	55073.1	54232.4	52925.0	52098.5	NaN	NaN	1377055.0	
4	2387656.4	2425290.6	2405613.8	2257604.3	NaN	NaN	60198837.5	
5	4559136.4	4620790.8	4661154.9	4621416.8	NaN	NaN	114506210.0	
6	131008.5	87233.6	93936.6	96103.4	NaN	NaN	2912167.5	
7	118003.0	118003.0	109463.8	110895.1	NaN	NaN	2864370.0	
8	326403.0	323700.3	322245.0	320581.8	NaN	NaN	8258827.5	
9	459003.1	473745.3	479362.7	468378.9	NaN	NaN	10920650.0	
10	163534.1	158489.7	154911.3	153260.9	NaN	NaN	4196555.0	
11	95630.0	93940.7	83333.6	77005.5	NaN	NaN	2370670.0	
12	121756.3	84456.3	111049.0	102156.3	NaN	NaN	3316000.0	
13	1052287.7	1087029.5	1118850.0	1083100.3	NaN	NaN	26485240.0	
14	123232.8	122656.3	125238.5	122453.8	NaN	NaN	3048565.0	
15	142723.7	162574.6	153890.8	156215.4	NaN	NaN	3731555.0	
16	381361.8	406083.3	393404.2	424755.1	NaN	NaN	9588232.5	
17	1524292.5	1609862.5	1606047.8	1556225.8	NaN	NaN	38945265.0	
18	83442.3	84596.2	83943.3	82937.0	NaN	NaN	2070582.5	
19	1703327.7	1714465.6	1702989.1	1553835.5	NaN	NaN	38856350.0	
20	49838.3	49991.2	49414.1	49152.0	NaN	NaN	1145967.5	
21	438662.0	336255.5	412177.5	350136.5	NaN	NaN	9118200.0	
22	1365173.8	1360551.6	1336231.2	1318431.1	NaN	NaN	34543610.0	

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	tCO2_2016	tCO2_2017	tCO2_2018	tCO2_2019	Mean_CH4	\
0	27337010.0	28863275.0	28614775.0	28618635.0	1133728.24	
1	3156955.0	3258072.5	3040380.0	2777120.0	125642.28	
2	134998000.0	135003225.0	132554327.5	130361367.5	5344654.14	
3	1376827.5	1355810.0	1323125.0	1302462.5	53882.24	
4	59691410.0	60632265.0	60140345.0	56440107.5	2376823.72	
5	113978410.0	115519770.0	116528872.5	115535420.0	4608549.46	
6	3275212.5	2180840.0	2348415.0	2402585.0	104953.76	
7	2950075.0	2950075.0	2736595.0	2772377.5	114187.94	
8	8160075.0	8092507.5	8056125.0	8014545.0	324656.64	
9	11475077.5	11843632.5	11984067.5	11709472.5	463463.20	
10	4088352.5	3962242.5	3872782.5	3831522.5	159611.64	
11	2390750.0	2348517.5	2083340.0	1925137.5	88947.32	
12	3043907.5	2111407.5	2776225.0	2553907.5	110411.58	
13	26307192.5	27175737.5	27971250.0	27077507.5	1080135.42	
14	3080820.0	3066407.5	3130962.5	3061345.0	123104.80	
15	3568092.5	4064365.0	3847270.0	3905385.0	152933.34	
16	9534045.0	10152082.5	9835105.0	10618877.5	397826.74	
17	38107312.5	40246562.5	40151195.0	38905645.0	1570847.84	
18	2086057.5	2114905.0	2098582.5	2073425.0	83548.42	
19	42583192.5	42861640.0	42574727.5	38845887.5	1645774.38	
20	1245957.5	1249780.0	1235352.5	1228800.0	48846.86	
21	10966550.0	8406387.5	10304437.5	8753412.5	380391.90	
22	34129345.0	34013790.0	33405780.0	32960777.5	1352426.42	

	Total_CH4	Mean_CO2	Total_CO2
0	5668641.2	28343206.0	141716030.0
1	628211.4	3141057.0	15705285.0
2	26723270.7	133616353.5	668081767.5
3	269411.2	1347056.0	6735280.0
4	11884118.6	59420593.0	297102965.0
5	23042747.3	115213736.5	576068682.5
6	524768.8	2623844.0	13119220.0
7	570939.7	2854698.5	14273492.5
8	1623283.2	8116416.0	40582080.0
9	2317316.0	11586580.0	57932900.0
10	798058.2	3990291.0	19951455.0
11	444736.6	2223683.0	11118415.0
12	552057.9	2760289.5	13801447.5
13	5400677.1	27003385.5	135016927.5
14	615524.0	3077620.0	15388100.0
15	764666.7	3823333.5	19116667.5
16	1989133.7	9945668.5	49728342.5
17	7854239.2	39271196.0	196355980.0
18	417742.1	2088710.5	10443552.5
19	8228871.9	41144359.5	205721797.5
20	244234.3	1221171.5	6105857.5
21	1901959.5	9509797.5	47548987.5
22	6762132.1	33810660.5	169053302.5

malaysia\_emissions\_df

iso3_country	country_name	tCH4_2015	\
0	BGD	Bangladesh	2.344420e+06
1	BRA	Brazil	3.410233e+05

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2	CHN	China	6.133647e+06
3	ESP	Spain	1.141464e+04
4	IDN	Indonesia	1.283649e+06
5	IND	India	6.219887e+06
6	IRN	Iran (Islamic Republic of)	8.774407e+04
7	ITA	Italy	4.995968e+04
8	JPN	Japan	2.305465e+05
9	KHM	Cambodia	4.954698e+05
10	KOR	Korea (the Republic of)	1.451878e+05
11	LAO	Lao People's Democratic Republic (the)	1.661169e+04
12	LKA	Sri Lanka	8.305626e+04
13	MMR	Myanmar	1.132082e+06
14	MYS	Malaysia	1.057399e+05
15	NPL	Nepal	1.007479e+05
16	PAK	Pakistan	4.852431e+05
17	PHL	Philippines (the)	3.432021e+05
18	PRK	Korea (the Democratic People's Republic of)	1.143217e+05
19	THA	Thailand	1.393798e+06
20	TWN	Taiwan (Province of China)	7.866956e+04
21	USA	United States of America (the)	1.611324e+05
22	VNM	Viet Nam	1.346013e+06

	tCH4_2016	tCH4_2017	tCH4_2018	tCH4_2019	tCH4_2020 \
0	2.278158e+06	2.098958e+06	2.141231e+06	2.070985e+06	2.106781e+06
1	3.104189e+05	3.725173e+05	3.717030e+05	3.294713e+05	4.902874e+05
2	5.859531e+06	6.355071e+06	5.413962e+06	5.603352e+06	6.402353e+06
3	1.334803e+04	1.217299e+04	1.405410e+04	1.148324e+04	1.305461e+04
4	1.023129e+06	9.615327e+05	1.176982e+06	1.266668e+06	1.188195e+06
5	5.309413e+06	6.228451e+06	6.589798e+06	7.501556e+06	7.599764e+06
6	9.180121e+04	9.620217e+04	8.875744e+04	9.500199e+04	9.600254e+04
7	4.937785e+04	5.443679e+04	4.469902e+04	4.566914e+04	5.101547e+04
8	2.284133e+05	2.708935e+05	1.548252e+05	2.332056e+05	2.835167e+05
9	5.731698e+05	4.517045e+05	5.592610e+05	5.947277e+05	6.412802e+05
10	1.274597e+05	1.463222e+05	1.293543e+05	1.327782e+05	1.165467e+05
11	1.696441e+04	1.168063e+04	1.009675e+04	1.461058e+04	2.136270e+04
12	1.011743e+05	5.911841e+04	9.018914e+04	8.476088e+04	9.248238e+04
13	1.290806e+06	1.205169e+06	1.372447e+06	1.256888e+06	1.221904e+06
14	1.110049e+05	1.111291e+05	1.066525e+05	1.056287e+05	1.127141e+05
15	6.667161e+04	8.081300e+04	9.200752e+04	1.164235e+05	7.168401e+04
16	5.945922e+05	5.372641e+05	4.532297e+05	6.528548e+05	6.401201e+05
17	4.073554e+05	3.836830e+05	4.175210e+05	3.584550e+05	4.462836e+05
18	9.177653e+04	1.085457e+05	8.662578e+04	9.655062e+04	8.581038e+04
19	1.780993e+06	1.164699e+06	9.166575e+05	1.305046e+06	1.520788e+06
20	8.089149e+04	8.705634e+04	8.138151e+04	8.990870e+04	8.333327e+04
21	1.618576e+05	1.684799e+05	1.657254e+05	1.691351e+05	1.941455e+05
22	1.483777e+06	1.406437e+06	1.317455e+06	1.269751e+06	1.374450e+06

	tCH4_2021	tCO2_2015	tCO2_2016	tCO2_2017	tCO2_2018 \
0	1.983974e+06	5.861049e+07	5.695395e+07	5.247394e+07	5.353076e+07
1	4.544874e+05	8.525583e+06	7.760473e+06	9.312934e+06	9.292575e+06
2	6.068210e+06	1.533412e+08	1.464883e+08	1.588768e+08	1.353491e+08
3	8.531579e+03	2.853661e+05	3.337007e+05	3.043248e+05	3.513524e+05
4	1.009936e+06	3.209122e+07	2.557824e+07	2.403832e+07	2.942454e+07
5	6.567960e+06	1.554972e+08	1.327353e+08	1.557113e+08	1.647450e+08
6	9.053525e+04	2.193602e+06	2.295030e+06	2.405054e+06	2.218936e+06
7	5.089759e+04	1.248992e+06	1.234446e+06	1.360920e+06	1.117475e+06

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8	1.574007e+05	5.763662e+06	5.710333e+06	6.772337e+06	3.870631e+06
9	5.644891e+05	1.238675e+07	1.432925e+07	1.129261e+07	1.398153e+07
10	1.013006e+05	3.629695e+06	3.186493e+06	3.658056e+06	3.233858e+06
11	1.475014e+04	4.152924e+05	4.241102e+05	2.920158e+05	2.524186e+05
12	8.466966e+04	2.076407e+06	2.529358e+06	1.477960e+06	2.254728e+06
13	1.289837e+06	2.830206e+07	3.227014e+07	3.012923e+07	3.431117e+07
14	1.069696e+05	2.643498e+06	2.775123e+06	2.778227e+06	2.666313e+06
15	4.811408e+04	2.518697e+06	1.666790e+06	2.020325e+06	2.300188e+06
16	4.849205e+05	1.213108e+07	1.486480e+07	1.343160e+07	1.133074e+07
17	4.383270e+05	8.580052e+06	1.018389e+07	9.592074e+06	1.043803e+07
18	7.735988e+04	2.858041e+06	2.294413e+06	2.713641e+06	2.165645e+06
19	8.528673e+05	3.484495e+07	4.452483e+07	2.911748e+07	2.291644e+07
20	6.619861e+04	1.966739e+06	2.022287e+06	2.176408e+06	2.034538e+06
21	1.634842e+05	4.028310e+06	4.046440e+06	4.211999e+06	4.143136e+06
22	1.502787e+06	3.365033e+07	3.709441e+07	3.516092e+07	3.293637e+07

	tCO2_2019	Mean_CH4	Total_CH4	Mean_CO2	Total_CO2
0	5.177463e+07	2.186750e+06	1.093375e+07	5.466875e+07	2.733438e+08
1	8.236783e+06	3.450268e+05	1.725134e+06	8.625670e+06	4.312835e+07
2	1.400838e+08	5.873113e+06	2.936556e+07	1.468278e+08	7.341391e+08
3	2.870810e+05	1.249460e+04	6.247300e+04	3.123650e+05	1.561825e+06
4	3.166670e+07	1.142392e+06	5.711960e+06	2.855980e+07	1.427990e+08
5	1.875389e+08	6.369821e+06	3.184910e+07	1.592455e+08	7.962276e+08
6	2.375050e+06	9.190138e+04	4.595069e+05	2.297534e+06	1.148767e+07
7	1.141729e+06	4.882850e+04	2.441425e+05	1.220712e+06	6.103562e+06
8	5.830141e+06	2.235768e+05	1.117884e+06	5.589421e+06	2.794710e+07
9	1.486819e+07	5.348666e+05	2.674333e+06	1.337166e+07	6.685832e+07
10	3.319455e+06	1.362205e+05	6.811023e+05	3.405512e+06	1.702756e+07
11	3.652645e+05	1.399281e+04	6.996406e+04	3.498203e+05	1.749102e+06
12	2.119022e+06	8.365981e+04	4.182990e+05	2.091495e+06	1.045748e+07
13	3.142221e+07	1.251478e+06	6.257392e+06	3.128696e+07	1.564348e+08
14	2.640717e+06	1.080310e+05	5.401551e+05	2.700775e+06	1.350388e+07
15	2.910588e+06	9.133271e+04	4.566635e+05	2.283318e+06	1.141659e+07
16	1.632137e+07	5.446368e+05	2.723184e+06	1.361592e+07	6.807960e+07
17	8.961374e+06	3.820433e+05	1.910216e+06	9.551082e+06	4.775541e+07
18	2.413765e+06	9.956405e+04	4.978202e+05	2.489101e+06	1.244551e+07
19	3.262615e+07	1.312239e+06	6.561194e+06	3.280597e+07	1.640298e+08
20	2.247717e+06	8.358152e+04	4.179076e+05	2.089538e+06	1.044769e+07
21	4.228377e+06	1.652661e+05	8.263305e+05	4.131652e+06	2.065826e+07
22	3.174377e+07	1.364686e+06	6.823432e+06	3.411716e+07	1.705858e+08

```
merged_df = faostat_emissions_df.merge(malaysia_emissions_df, suffixes=('_FAOSTAT', '_
↳TRACE'), on='iso3_country', how='left', sort=False)
```

## 5.7.2 Dropping 2020 and 2021 from the data sets

I will only compare data compiled from the same year.

```
merged_df.drop([2020, 2021, "tCH4_2020", "tCH4_2021"], axis = 1, inplace = True)
```

```
merged_df
```

	iso3_country	country_name_FAOSTAT	2015	\
0	BGD	Bangladesh	1131293.4	
1	BRA	Brazil	138910.3	
2	CHN	China	5406593.9	
3	ESP	Spain	55082.2	
4	IDN	Indonesia	2407953.5	
5	IND	India	4580248.4	
6	IRN	Iran (Islamic Republic of)	116486.7	
7	ITA	Italy	114574.8	
8	JPN	Japan	330353.1	
9	KHM	Cambodia	436826.0	
10	KOR	Korea (the Republic of)	167862.2	
11	LAO	Lao People's Democratic Republic (the)	94826.8	
12	LKA	Sri Lanka	132640.0	
13	MMR	Myanmar	1059409.6	
14	MYS	Malaysia	121942.6	
15	NPL	Nepal	149262.2	
16	PAK	Pakistan	383529.3	
17	PHL	Philippines (the)	1557810.6	
18	PRK	Korea (the Democratic People's Republic of)	82823.3	
19	THA	Thailand	1554254.0	
20	TWN	Taiwan (Province of China)	45838.7	
21	USA	United States of America (the)	364728.0	
22	VNM	Viet Nam	1381744.4	

	2016	2017	2018	2019	tCO2_2015_FAOSTAT	\
0	1093480.4	1154531.0	1144591.0	1144745.4	28282335.0	
1	126278.2	130322.9	121615.2	111084.8	3472757.5	
2	5399920.0	5400129.0	5302173.1	5214454.7	135164847.5	
3	55073.1	54232.4	52925.0	52098.5	1377055.0	
4	2387656.4	2425290.6	2405613.8	2257604.3	60198837.5	
5	4559136.4	4620790.8	4661154.9	4621416.8	114506210.0	
6	131008.5	87233.6	93936.6	96103.4	2912167.5	
7	118003.0	118003.0	109463.8	110895.1	2864370.0	
8	326403.0	323700.3	322245.0	320581.8	8258827.5	
9	459003.1	473745.3	479362.7	468378.9	10920650.0	
10	163534.1	158489.7	154911.3	153260.9	4196555.0	
11	95630.0	93940.7	83333.6	77005.5	2370670.0	
12	121756.3	84456.3	111049.0	102156.3	3316000.0	
13	1052287.7	1087029.5	1118850.0	1083100.3	26485240.0	
14	123232.8	122656.3	125238.5	122453.8	3048565.0	
15	142723.7	162574.6	153890.8	156215.4	3731555.0	
16	381361.8	406083.3	393404.2	424755.1	9588232.5	
17	1524292.5	1609862.5	1606047.8	1556225.8	38945265.0	
18	83442.3	84596.2	83943.3	82937.0	2070582.5	
19	1703327.7	1714465.6	1702989.1	1553835.5	38856350.0	
20	49838.3	49991.2	49414.1	49152.0	1145967.5	
21	438662.0	336255.5	412177.5	350136.5	9118200.0	
22	1365173.8	1360551.6	1336231.2	1318431.1	34543610.0	

	tCO2_2016_FAOSTAT	tCO2_2017_FAOSTAT	...	tCH4_2019	tCO2_2015_TRACE	\
0	27337010.0	28863275.0	...	2.070985e+06	5.861049e+07	
1	3156955.0	3258072.5	...	3.294713e+05	8.525583e+06	
2	134998000.0	135003225.0	...	5.603352e+06	1.533412e+08	
3	1376827.5	1355810.0	...	1.148324e+04	2.853661e+05	
4	59691410.0	60632265.0	...	1.266668e+06	3.209122e+07	
5	113978410.0	115519770.0	...	7.501556e+06	1.554972e+08	

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6	3275212.5	2180840.0	...	9.500199e+04	2.193602e+06
7	2950075.0	2950075.0	...	4.566914e+04	1.248992e+06
8	8160075.0	8092507.5	...	2.332056e+05	5.763662e+06
9	11475077.5	11843632.5	...	5.947277e+05	1.238675e+07
10	4088352.5	3962242.5	...	1.327782e+05	3.629695e+06
11	2390750.0	2348517.5	...	1.461058e+04	4.152924e+05
12	3043907.5	2111407.5	...	8.476088e+04	2.076407e+06
13	26307192.5	27175737.5	...	1.256888e+06	2.830206e+07
14	3080820.0	3066407.5	...	1.056287e+05	2.643498e+06
15	3568092.5	4064365.0	...	1.164235e+05	2.518697e+06
16	9534045.0	10152082.5	...	6.528548e+05	1.213108e+07
17	38107312.5	40246562.5	...	3.584550e+05	8.580052e+06
18	2086057.5	2114905.0	...	9.655062e+04	2.858041e+06
19	42583192.5	42861640.0	...	1.305046e+06	3.484495e+07
20	1245957.5	1249780.0	...	8.990870e+04	1.966739e+06
21	10966550.0	8406387.5	...	1.691351e+05	4.028310e+06
22	34129345.0	34013790.0	...	1.269751e+06	3.365033e+07

	tCO2_2016_TRACE	tCO2_2017_TRACE	tCO2_2018_TRACE	tCO2_2019_TRACE	\
0	5.695395e+07	5.247394e+07	5.353076e+07	5.177463e+07	
1	7.760473e+06	9.312934e+06	9.292575e+06	8.236783e+06	
2	1.464883e+08	1.588768e+08	1.353491e+08	1.400838e+08	
3	3.337007e+05	3.043248e+05	3.513524e+05	2.870810e+05	
4	2.557824e+07	2.403832e+07	2.942454e+07	3.166670e+07	
5	1.327353e+08	1.557113e+08	1.647450e+08	1.875389e+08	
6	2.295030e+06	2.405054e+06	2.218936e+06	2.375050e+06	
7	1.234446e+06	1.360920e+06	1.117475e+06	1.141729e+06	
8	5.710333e+06	6.772337e+06	3.870631e+06	5.830141e+06	
9	1.432925e+07	1.129261e+07	1.398153e+07	1.486819e+07	
10	3.186493e+06	3.658056e+06	3.233858e+06	3.319455e+06	
11	4.241102e+05	2.920158e+05	2.524186e+05	3.652645e+05	
12	2.529358e+06	1.477960e+06	2.254728e+06	2.119022e+06	
13	3.227014e+07	3.012923e+07	3.431117e+07	3.142221e+07	
14	2.775123e+06	2.778227e+06	2.666313e+06	2.640717e+06	
15	1.666790e+06	2.020325e+06	2.300188e+06	2.910588e+06	
16	1.486480e+07	1.343160e+07	1.133074e+07	1.632137e+07	
17	1.018389e+07	9.592074e+06	1.043803e+07	8.961374e+06	
18	2.294413e+06	2.713641e+06	2.165645e+06	2.413765e+06	
19	4.452483e+07	2.911748e+07	2.291644e+07	3.262615e+07	
20	2.022287e+06	2.176408e+06	2.034538e+06	2.247717e+06	
21	4.046440e+06	4.211999e+06	4.143136e+06	4.228377e+06	
22	3.709441e+07	3.516092e+07	3.293637e+07	3.174377e+07	

	Mean_CH4_TRACE	Total_CH4_TRACE	Mean_CO2_TRACE	Total_CO2_TRACE
0	2.186750e+06	1.093375e+07	5.466875e+07	2.733438e+08
1	3.450268e+05	1.725134e+06	8.625670e+06	4.312835e+07
2	5.873113e+06	2.936556e+07	1.468278e+08	7.341391e+08
3	1.249460e+04	6.247300e+04	3.123650e+05	1.561825e+06
4	1.142392e+06	5.711960e+06	2.855980e+07	1.427990e+08
5	6.369821e+06	3.184910e+07	1.592455e+08	7.962276e+08
6	9.190138e+04	4.595069e+05	2.297534e+06	1.148767e+07
7	4.882850e+04	2.441425e+05	1.220712e+06	6.103562e+06
8	2.235768e+05	1.117884e+06	5.589421e+06	2.794710e+07
9	5.348666e+05	2.674333e+06	1.337166e+07	6.685832e+07
10	1.362205e+05	6.811023e+05	3.405512e+06	1.702756e+07
11	1.399281e+04	6.996406e+04	3.498203e+05	1.749102e+06

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```

12  8.365981e+04  4.182990e+05  2.091495e+06  1.045748e+07
13  1.251478e+06  6.257392e+06  3.128696e+07  1.564348e+08
14  1.080310e+05  5.401551e+05  2.700775e+06  1.350388e+07
15  9.133271e+04  4.566635e+05  2.283318e+06  1.141659e+07
16  5.446368e+05  2.723184e+06  1.361592e+07  6.807960e+07
17  3.820433e+05  1.910216e+06  9.551082e+06  4.775541e+07
18  9.956405e+04  4.978202e+05  2.489101e+06  1.244551e+07
19  1.312239e+06  6.561194e+06  3.280597e+07  1.640298e+08
20  8.358152e+04  4.179076e+05  2.089538e+06  1.044769e+07
21  1.652661e+05  8.263305e+05  4.131652e+06  2.065826e+07
22  1.364686e+06  6.823432e+06  3.411716e+07  1.705858e+08

```

[23 rows x 31 columns]

### 5.7.3 Calculate difference in Ch4 Tonnes Between the Estimates

```

# Calculate Difference in tons
merged_df['CH4_diff_2015'] = merged_df[2015] - merged_df['tCH4_2015']
merged_df['CH4_diff_2016'] = merged_df[2016] - merged_df['tCH4_2016']
merged_df['CH4_diff_2017'] = merged_df[2017] - merged_df['tCH4_2017']
merged_df['CH4_diff_2018'] = merged_df[2018] - merged_df['tCH4_2018']
merged_df['CH4_diff_2019'] = merged_df[2019] - merged_df['tCH4_2019']
merged_df['CH4_diff_means'] = merged_df['Mean_CH4_FAOSTAT'] - merged_df['Mean_CH4_
↳TRACE']
merged_df['CH4_diff_totals'] = merged_df['Total_CH4_FAOSTAT'] - merged_df['Total_CH4_
↳TRACE']

```

merged\_df

	iso3_country	country_name_FAOSTAT	2015 \
0	BGD	Bangladesh	1131293.4
1	BRA	Brazil	138910.3
2	CHN	China	5406593.9
3	ESP	Spain	55082.2
4	IDN	Indonesia	2407953.5
5	IND	India	4580248.4
6	IRN	Iran (Islamic Republic of)	116486.7
7	ITA	Italy	114574.8
8	JPN	Japan	330353.1
9	KHM	Cambodia	436826.0
10	KOR	Korea (the Republic of)	167862.2
11	LAO	Lao People's Democratic Republic (the)	94826.8
12	LKA	Sri Lanka	132640.0
13	MMR	Myanmar	1059409.6
14	MYS	Malaysia	121942.6
15	NPL	Nepal	149262.2
16	PAK	Pakistan	383529.3
17	PHL	Philippines (the)	1557810.6
18	PRK	Korea (the Democratic People's Republic of)	82823.3
19	THA	Thailand	1554254.0
20	TWN	Taiwan (Province of China)	45838.7
21	USA	United States of America (the)	364728.0

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22	VNM				Viet Nam 1381744.4	
	2016	2017	2018	2019	tCO2_2015_FAOSTAT	\
0	1093480.4	1154531.0	1144591.0	1144745.4	28282335.0	
1	126278.2	130322.9	121615.2	111084.8	3472757.5	
2	5399920.0	5400129.0	5302173.1	5214454.7	135164847.5	
3	55073.1	54232.4	52925.0	52098.5	1377055.0	
4	2387656.4	2425290.6	2405613.8	2257604.3	60198837.5	
5	4559136.4	4620790.8	4661154.9	4621416.8	114506210.0	
6	131008.5	87233.6	93936.6	96103.4	2912167.5	
7	118003.0	118003.0	109463.8	110895.1	2864370.0	
8	326403.0	323700.3	322245.0	320581.8	8258827.5	
9	459003.1	473745.3	479362.7	468378.9	10920650.0	
10	163534.1	158489.7	154911.3	153260.9	4196555.0	
11	95630.0	93940.7	83333.6	77005.5	2370670.0	
12	121756.3	84456.3	111049.0	102156.3	3316000.0	
13	1052287.7	1087029.5	1118850.0	1083100.3	26485240.0	
14	123232.8	122656.3	125238.5	122453.8	3048565.0	
15	142723.7	162574.6	153890.8	156215.4	3731555.0	
16	381361.8	406083.3	393404.2	424755.1	9588232.5	
17	1524292.5	1609862.5	1606047.8	1556225.8	38945265.0	
18	83442.3	84596.2	83943.3	82937.0	2070582.5	
19	1703327.7	1714465.6	1702989.1	1553835.5	38856350.0	
20	49838.3	49991.2	49414.1	49152.0	1145967.5	
21	438662.0	336255.5	412177.5	350136.5	9118200.0	
22	1365173.8	1360551.6	1336231.2	1318431.1	34543610.0	
	tCO2_2016_FAOSTAT	tCO2_2017_FAOSTAT	...	Total_CH4_TRACE	\	
0	27337010.0	28863275.0	...	1.093375e+07		
1	3156955.0	3258072.5	...	1.725134e+06		
2	134998000.0	135003225.0	...	2.936556e+07		
3	1376827.5	1355810.0	...	6.247300e+04		
4	59691410.0	60632265.0	...	5.711960e+06		
5	113978410.0	115519770.0	...	3.184910e+07		
6	3275212.5	2180840.0	...	4.595069e+05		
7	2950075.0	2950075.0	...	2.441425e+05		
8	8160075.0	8092507.5	...	1.117884e+06		
9	11475077.5	11843632.5	...	2.674333e+06		
10	4088352.5	3962242.5	...	6.811023e+05		
11	2390750.0	2348517.5	...	6.996406e+04		
12	3043907.5	2111407.5	...	4.182990e+05		
13	26307192.5	27175737.5	...	6.257392e+06		
14	3080820.0	3066407.5	...	5.401551e+05		
15	3568092.5	4064365.0	...	4.566635e+05		
16	9534045.0	10152082.5	...	2.723184e+06		
17	38107312.5	40246562.5	...	1.910216e+06		
18	2086057.5	2114905.0	...	4.978202e+05		
19	42583192.5	42861640.0	...	6.561194e+06		
20	1245957.5	1249780.0	...	4.179076e+05		
21	10966550.0	8406387.5	...	8.263305e+05		
22	34129345.0	34013790.0	...	6.823432e+06		
	Mean_CO2_TRACE	Total_CO2_TRACE	CH4_diff_2015	CH4_diff_2016	\	
0	5.466875e+07	2.733438e+08	-1.213126e+06	-1.184678e+06		
1	8.625670e+06	4.312835e+07	-2.021130e+05	-1.841407e+05		
2	1.468278e+08	7.341391e+08	-7.270532e+05	-4.596111e+05		

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3	3.123650e+05	1.561825e+06	4.366756e+04	4.172507e+04
4	2.855980e+07	1.427990e+08	1.124305e+06	1.364527e+06
5	1.592455e+08	7.962276e+08	-1.639638e+06	-7.502765e+05
6	2.297534e+06	1.148767e+07	2.874263e+04	3.920729e+04
7	1.220712e+06	6.103562e+06	6.461512e+04	6.862515e+04
8	5.589421e+06	2.794710e+07	9.980661e+04	9.798967e+04
9	1.337166e+07	6.685832e+07	-5.864384e+04	-1.141667e+05
10	3.405512e+06	1.702756e+07	2.267438e+04	3.607436e+04
11	3.498203e+05	1.749102e+06	7.821511e+04	7.866559e+04
12	2.091495e+06	1.045748e+07	4.958374e+04	2.058196e+04
13	3.128696e+07	1.564348e+08	-7.267265e+04	-2.385180e+05
14	2.700775e+06	1.350388e+07	1.620268e+04	1.222789e+04
15	2.283318e+06	1.141659e+07	4.851434e+04	7.605209e+04
16	1.361592e+07	6.807960e+07	-1.017138e+05	-2.132304e+05
17	9.551082e+06	4.775541e+07	1.214609e+06	1.116937e+06
18	2.489101e+06	1.244551e+07	-3.149836e+04	-8.334231e+03
19	3.280597e+07	1.640298e+08	1.604559e+05	-7.766539e+04
20	2.089538e+06	1.044769e+07	-3.283086e+04	-3.105319e+04
21	4.131652e+06	2.065826e+07	2.035956e+05	2.768044e+05
22	3.411716e+07	1.705858e+08	3.573139e+04	-1.186027e+05

	CH4_diff_2017	CH4_diff_2018	CH4_diff_2019	CH4_diff_means \
0	-9.444266e+05	-9.966395e+05	-9.262397e+05	-1.053022e+06
1	-2.421944e+05	-2.500878e+05	-2.183865e+05	-2.193845e+05
2	-9.549420e+05	-1.117892e+05	-3.888972e+05	-5.284585e+05
3	4.205941e+04	3.887090e+04	4.061526e+04	4.138764e+04
4	1.463758e+06	1.228632e+06	9.909364e+05	1.234432e+06
5	-1.607660e+06	-1.928643e+06	-2.880139e+06	-1.761271e+06
6	-8.968572e+03	5.179159e+03	1.101412e+03	1.305238e+04
7	6.356621e+04	6.476478e+04	6.522596e+04	6.535944e+04
8	5.280682e+04	1.674198e+05	8.737617e+04	1.010798e+05
9	2.204082e+04	-7.989830e+04	-1.263488e+05	-7.140337e+04
10	1.216747e+04	2.555698e+04	2.048269e+04	2.339118e+04
11	8.226007e+04	7.323685e+04	6.239492e+04	7.495451e+04
12	2.533789e+04	2.085986e+04	1.739542e+04	2.675177e+04
13	-1.181398e+05	-2.535969e+05	-1.737880e+05	-1.713431e+05
14	1.152724e+04	1.858597e+04	1.682512e+04	1.507378e+04
15	8.176160e+04	6.188328e+04	3.979186e+04	6.160063e+04
16	-1.311808e+05	-5.982552e+04	-2.280997e+05	-1.468100e+05
17	1.226180e+06	1.188527e+06	1.197771e+06	1.188805e+06
18	-2.394946e+04	-2.682481e+03	-1.361362e+04	-1.601563e+04
19	5.497665e+05	7.863316e+05	2.487897e+05	3.335357e+05
20	-3.706514e+04	-3.196741e+04	-4.075670e+04	-3.473466e+04
21	1.677756e+05	2.464521e+05	1.810014e+05	2.151258e+05
22	-4.588520e+04	1.877621e+04	4.868034e+04	-1.226000e+04

	CH4_diff_totals
0	-5.265110e+06
1	-1.096923e+06
2	-2.642293e+06
3	2.069382e+05
4	6.172158e+06
5	-8.806357e+06
6	6.526191e+04
7	3.267972e+05
8	5.053990e+05

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```

9      -3.570168e+05
10     1.169559e+05
11     3.747725e+05
12     1.337589e+05
13    -8.567154e+05
14     7.536891e+04
15     3.080032e+05
16    -7.340502e+05
17     5.944023e+06
18    -8.007815e+04
19     1.667678e+06
20    -1.736733e+05
21     1.075629e+06
22    -6.129998e+04

[23 rows x 38 columns]
```

## 5.7.4 Calculate difference in CO2 Tonnes Between the Estimates

```

# Calculate Difference in tons
merged_df['CO2_diff_2015'] = merged_df['tCO2_2015_FAOSTAT'] - merged_df['tCO2_2015_
↳TRACE']
merged_df['CO2_diff_2016'] = merged_df['tCO2_2016_FAOSTAT'] - merged_df['tCO2_2016_
↳TRACE']
merged_df['CO2_diff_2017'] = merged_df['tCO2_2017_FAOSTAT'] - merged_df['tCO2_2017_
↳TRACE']
merged_df['CO2_diff_2018'] = merged_df['tCO2_2018_FAOSTAT'] - merged_df['tCO2_2018_
↳TRACE']
merged_df['CO2_diff_2019'] = merged_df['tCO2_2019_FAOSTAT'] - merged_df['tCO2_2019_
↳TRACE']
merged_df['CO2_diff_means'] = merged_df['Mean_CO2_FAOSTAT'] - merged_df['Mean_CO2_
↳TRACE']
merged_df['CO2_diff_totals'] = merged_df['Total_CO2_FAOSTAT'] - merged_df['Total_CO2_
↳TRACE']
```

## 5.7.5 Calculating the CH4 Percent Differences Between the Estimates

```

## Calculate Percent Differences on this data set )*100
# With raw data i could have accomplished this with a groupby.aggregate(lambda x ),
↳however the pivot tables given are not easy to apply #vectorized functions across
↳time series
merged_df['CH4_abs_percent_diff_2015'] = ((abs(merged_df[2015] - merged_df['tCH4_2015
↳'])) / ((merged_df[2015] + merged_df['tCH4_2015']) / 2)) * 100
merged_df['CH4_abs_percent_diff_2016'] = ((abs(merged_df[2016] - merged_df['tCH4_2016
↳'])) / ((merged_df[2016] + merged_df['tCH4_2016']) / 2)) * 100
merged_df['CH4_abs_percent_diff_2017'] = ((abs(merged_df[2017] - merged_df['tCH4_2017
↳'])) / ((merged_df[2017] + merged_df['tCH4_2017']) / 2)) * 100
merged_df['CH4_abs_percent_diff_2018'] = (abs((merged_df[2018] - merged_df['tCH4_2018
↳'])) / ((merged_df[2018] + merged_df['tCH4_2018']) / 2)) * 100
merged_df['CH4_abs_percent_diff_2019'] = (abs((merged_df[2019] - merged_df['tCH4_2019
↳'])) / ((merged_df[2019] + merged_df['tCH4_2019']) / 2)) * 100
```

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```
merged_df['CH4_abs_percent_diff_means'] = (abs((merged_df['Mean_CH4_FAOSTAT'] -
↪merged_df['Mean_CH4_TRACE']))/((merged_df['Mean_CH4_FAOSTAT'] + merged_df['Mean_CH4_
↪TRACE'])/2))*100
merged_df['CH4_abs_percent_diff_totals'] = (abs((merged_df['Total_CH4_FAOSTAT'] -
↪merged_df['Total_CH4_TRACE']))/((merged_df['Total_CH4_TRACE'] + merged_df['Total_
↪CH4_FAOSTAT'])/2))*100

merged_df['CH4_relative_percent_diff_2015'] = ((merged_df[2015] - merged_df['tCH4_2015
↪'])/(merged_df[2015]))*100
merged_df['CH4_relative_percent_diff_2016'] = ((merged_df[2016] - merged_df['tCH4_2016
↪'])/(merged_df[2016]))*100
merged_df['CH4_relative_percent_diff_2017'] = ((merged_df[2017] - merged_df['tCH4_2017
↪'])/(merged_df[2017]))*100
merged_df['CH4_relative_percent_diff_2018'] = ((merged_df[2018] - merged_df['tCH4_2018
↪'])/(merged_df[2018]))*100
merged_df['CH4_relative_percent_diff_2019'] = ((merged_df[2019] - merged_df['tCH4_2019
↪'])/(merged_df[2019]))*100
merged_df['CH4_relative_percent_diff_means'] = ((merged_df['Mean_CH4_FAOSTAT'] -
↪merged_df['Mean_CH4_TRACE'])/(merged_df["Mean_CH4_FAOSTAT"]))*100
merged_df['CH4_relative_percent_diff_totals'] = ((merged_df['Total_CH4_FAOSTAT'] -
↪merged_df['Total_CH4_TRACE'])/(merged_df["Total_CH4_FAOSTAT"]))*100
```

## 5.7.6 Calculate CO2 Differences

```
## Calculate Percent Differences on this data set )*100
# With raw data i could have accomplished this with a groupby.aggregate(lambda x ),
↪however the pivot tables given are not easy to apply #vectorized functions across
↪time series
merged_df['CO2_abs_percent_diff_2015'] = (abs((merged_df['tCO2_2015_FAOSTAT'] -
↪merged_df['tCO2_2015_TRACE']))/((merged_df['tCO2_2015_TRACE'] + merged_df['tCO2_
↪2015_FAOSTAT'])/2))*100
merged_df['CO2_abs_percent_diff_2016'] = ((abs(merged_df['tCO2_2016_FAOSTAT'] -
↪merged_df['tCO2_2016_TRACE']))/((merged_df['tCO2_2016_TRACE'] + merged_df['tCO2_
↪2016_FAOSTAT'])/2))*100
merged_df['CO2_abs_percent_diff_2017'] = ((abs(merged_df['tCO2_2017_FAOSTAT'] -
↪merged_df['tCO2_2017_TRACE']))/((merged_df['tCO2_2017_TRACE'] + merged_df['tCO2_
↪2017_FAOSTAT'])/2))*100
merged_df['CO2_abs_percent_diff_2018'] = ((abs(merged_df['tCO2_2018_FAOSTAT'] -
↪merged_df['tCO2_2018_TRACE']))/((merged_df['tCO2_2018_TRACE'] + merged_df['tCO2_
↪2018_FAOSTAT'])/2))*100
merged_df['CO2_abs_percent_diff_2019'] = ((abs(merged_df['tCO2_2019_FAOSTAT'] -
↪merged_df['tCO2_2019_TRACE']))/((merged_df['tCO2_2019_TRACE'] + merged_df['tCO2_
↪2019_FAOSTAT'])/2))*100
merged_df['CO2_abs_percent_diff_means'] = ((abs(merged_df['Mean_CO2_FAOSTAT'] -
↪merged_df['Mean_CO2_TRACE']))/((merged_df['Mean_CO2_FAOSTAT'] + merged_df['Mean_CO2_
↪TRACE'])/2))*100
merged_df['CO2_abs_percent_diff_totals'] = ((abs(merged_df['Total_CO2_FAOSTAT'] -
↪merged_df['Total_CO2_TRACE']))/((merged_df['Total_CO2_TRACE'] + merged_df['Total_
↪CO2_FAOSTAT'])/2))*100

merged_df['CO2_relative_percent_diff_2015'] = ((merged_df['tCO2_2015_FAOSTAT'] -
↪merged_df['tCO2_2015_TRACE'])/(merged_df['tCO2_2015_FAOSTAT']))*100
```

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```
merged_df['CO2_relative_percent_diff_2016'] = ((merged_df['tCO2_2016_FAOSTAT'] -
↳merged_df['tCO2_2016_TRACE'])/(merged_df['tCO2_2016_FAOSTAT']))*100
merged_df['CO2_relative_percent_diff_2017'] = ((merged_df['tCO2_2017_FAOSTAT'] -
↳merged_df['tCO2_2017_TRACE'])/(merged_df['tCO2_2017_FAOSTAT']))*100
merged_df['CO2_relative_percent_diff_2018'] = ((merged_df['tCO2_2018_FAOSTAT'] -
↳merged_df['tCO2_2018_TRACE'])/(merged_df['tCO2_2018_FAOSTAT']))*100
merged_df['CO2_relative_percent_diff_2019'] = ((merged_df['tCO2_2019_FAOSTAT'] -
↳merged_df['tCO2_2019_TRACE'])/(merged_df['tCO2_2019_FAOSTAT']))*100
merged_df['CO2_relative_percent_diff_means'] = ((merged_df["Mean_CO2_FAOSTAT"] -
↳merged_df["Mean_CO2_TRACE"])/(merged_df["Mean_CO2_FAOSTAT"]))*100
merged_df['CO2_relative_percent_diff_totals'] = ((merged_df['Total_CO2_FAOSTAT'] -
↳merged_df['Total_CO2_TRACE'])/(merged_df["Total_CO2_FAOSTAT"]))*100
```

merged\_df

	iso3_country	country_name_FAOSTAT	2015	\
0	BGD	Bangladesh	1131293.4	
1	BRA	Brazil	138910.3	
2	CHN	China	5406593.9	
3	ESP	Spain	55082.2	
4	IDN	Indonesia	2407953.5	
5	IND	India	4580248.4	
6	IRN	Iran (Islamic Republic of)	116486.7	
7	ITA	Italy	114574.8	
8	JPN	Japan	330353.1	
9	KHM	Cambodia	436826.0	
10	KOR	Korea (the Republic of)	167862.2	
11	LAO	Lao People's Democratic Republic (the)	94826.8	
12	LKA	Sri Lanka	132640.0	
13	MMR	Myanmar	1059409.6	
14	MYS	Malaysia	121942.6	
15	NPL	Nepal	149262.2	
16	PAK	Pakistan	383529.3	
17	PHL	Philippines (the)	1557810.6	
18	PRK	Korea (the Democratic People's Republic of)	82823.3	
19	THA	Thailand	1554254.0	
20	TWN	Taiwan (Province of China)	45838.7	
21	USA	United States of America (the)	364728.0	
22	VNM	Viet Nam	1381744.4	

	2016	2017	2018	2019	tCO2_2015_FAOSTAT	\
0	1093480.4	1154531.0	1144591.0	1144745.4	28282335.0	
1	126278.2	130322.9	121615.2	111084.8	3472757.5	
2	5399920.0	5400129.0	5302173.1	5214454.7	135164847.5	
3	55073.1	54232.4	52925.0	52098.5	1377055.0	
4	2387656.4	2425290.6	2405613.8	2257604.3	60198837.5	
5	4559136.4	4620790.8	4661154.9	4621416.8	114506210.0	
6	131008.5	87233.6	93936.6	96103.4	2912167.5	
7	118003.0	118003.0	109463.8	110895.1	2864370.0	
8	326403.0	323700.3	322245.0	320581.8	8258827.5	
9	459003.1	473745.3	479362.7	468378.9	10920650.0	
10	163534.1	158489.7	154911.3	153260.9	4196555.0	
11	95630.0	93940.7	83333.6	77005.5	2370670.0	
12	121756.3	84456.3	111049.0	102156.3	3316000.0	
13	1052287.7	1087029.5	1118850.0	1083100.3	26485240.0	

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14	123232.8	122656.3	125238.5	122453.8	3048565.0
15	142723.7	162574.6	153890.8	156215.4	3731555.0
16	381361.8	406083.3	393404.2	424755.1	9588232.5
17	1524292.5	1609862.5	1606047.8	1556225.8	38945265.0
18	83442.3	84596.2	83943.3	82937.0	2070582.5
19	1703327.7	1714465.6	1702989.1	1553835.5	38856350.0
20	49838.3	49991.2	49414.1	49152.0	1145967.5
21	438662.0	336255.5	412177.5	350136.5	9118200.0
22	1365173.8	1360551.6	1336231.2	1318431.1	34543610.0
	tCO2_2016_FAOSTAT	tCO2_2017_FAOSTAT	...	CO2_abs_percent_diff_2019	\
0	27337010.0	28863275.0	...	57.606797	
1	3156955.0	3258072.5	...	99.141295	
2	134998000.0	135003225.0	...	7.189945	
3	1376827.5	1355810.0	...	127.757634	
4	59691410.0	60632265.0	...	56.234955	
5	113978410.0	115519770.0	...	47.515394	
6	3275212.5	2180840.0	...	1.152675	
7	2950075.0	2950075.0	...	83.321649	
8	8160075.0	8092507.5	...	31.555853	
9	11475077.5	11843632.5	...	23.769728	
10	4088352.5	3962242.5	...	14.321600	
11	2390750.0	2348517.5	...	136.209544	
12	3043907.5	2111407.5	...	18.612972	
13	26307192.5	27175737.5	...	14.853751	
14	3080820.0	3066407.5	...	14.753544	
15	3568092.5	4064365.0	...	29.190154	
16	9534045.0	10152082.5	...	42.334367	
17	38107312.5	40246562.5	...	125.114417	
18	2086057.5	2114905.0	...	15.169422	
19	42583192.5	42861640.0	...	17.404685	
20	1245957.5	1249780.0	...	58.617135	
21	10966550.0	8406387.5	...	69.713577	
22	34129345.0	34013790.0	...	3.761740	
	CO2_abs_percent_diff_means	CO2_abs_percent_diff_totals	\		
0	63.425917	63.425917			
1	93.222402	93.222402			
2	9.421813	9.421813			
3	124.705066	124.705066			
4	70.153790	70.153790			
5	32.086209	32.086209			
6	13.260902	13.260902			
7	80.187551	80.187551			
8	36.874729	36.874729			
9	14.304565	14.304565			
10	15.813819	15.813819			
11	145.627378	145.627378			
12	27.569003	27.569003			
13	14.697380	14.697380			
14	13.043224	13.043224			
15	50.437328	50.437328			
16	31.154530	31.154530			
17	121.748165	121.748165			
18	17.492669	17.492669			
19	22.551331	22.551331			

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20	52.458032	52.458032
21	78.850049	78.850049
22	0.902428	0.902428
	CO2_relative_percent_diff_2015	CO2_relative_percent_diff_2016 \
0	-107.233570	-108.340089
1	-145.498958	-145.821472
2	-13.447527	-8.511442
3	79.277072	75.763074
4	46.691296	57.149216
5	-35.798018	-16.456550
6	24.674600	29.927285
7	56.395576	58.155426
8	30.212099	30.021068
9	-13.424988	-24.872761
10	13.507737	22.059230
11	82.482067	82.260369
12	37.382189	16.904226
13	-6.859731	-22.666619
14	13.287137	9.922595
15	32.502761	53.286238
16	-26.520489	-55.912882
17	77.968948	73.275770
18	-38.030793	-9.988016
19	10.323662	-4.559627
20	-71.622589	-62.307893
21	55.821215	63.101978
22	2.585962	-8.687738
	CO2_relative_percent_diff_2017	CO2_relative_percent_diff_2018 \
0	-81.801753	-87.073857
1	-185.841820	-205.638590
2	-17.683689	-2.108366
3	77.554025	73.445258
4	60.353916	51.073546
5	-34.791882	-41.376939
6	-10.281098	5.513463
7	53.868303	59.165480
8	16.313490	51.954183
9	4.652461	-16.667610
10	7.677136	16.497815
11	87.565950	87.883944
12	30.001182	18.784377
13	-10.868129	-22.665852
14	9.398001	14.840462
15	50.291743	40.212459
16	-32.303912	-15.207139
17	76.166725	74.003201
18	-28.310326	-3.195586
19	32.066348	46.173612
20	-74.143329	-64.692890
21	49.895259	59.792699
22	-3.372544	1.405162
	CO2_relative_percent_diff_2019	CO2_relative_percent_diff_means \
0	-80.912284	-92.881337

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1	-196.594424	-174.610414
2	-7.458060	-9.887609
3	77.958601	76.811284
4	43.893273	51.936188
5	-62.321563	-38.217479
6	1.146070	12.436318
7	58.817707	57.238482
8	27.255501	31.134372
9	-26.975760	-15.406481
10	13.364589	14.655057
11	81.026578	84.268427
12	17.028241	24.229137
13	-16.045425	-15.863111
14	13.739977	12.244674
15	25.472433	40.279401
16	-53.701452	-36.903009
17	76.966391	75.679166
18	-16.414407	-19.169278
19	16.011325	20.266184
20	-82.919716	-71.109302
21	51.694526	56.553728
22	3.692293	-0.906518

	CO2_relative_percent_diff_totals
0	-92.881337
1	-174.610414
2	-9.887609
3	76.811284
4	51.936188
5	-38.217479
6	12.436318
7	57.238482
8	31.134372
9	-15.406481
10	14.655057
11	84.268427
12	24.229137
13	-15.863111
14	12.244674
15	40.279401
16	-36.903009
17	75.679166
18	-19.169278
19	20.266184
20	-71.109302
21	56.553728
22	-0.906518

[23 rows x 73 columns]

## 5.7.7 Recalculate Means

```
merged_df.loc['mean'] = merged_df.select_dtypes(np.number).mean()
```

```
merged_df.at['mean', 'country_name_FAOSTAT'] = 'mean'
merged_df.at['mean', 'country_name_TRACE'] = 'mean'
```

```
merged_df
```

	iso3_country	country_name_FAOSTAT	2015	\
0	BGD	Bangladesh	1131293.4	
1	BRA	Brazil	138910.3	
2	CHN	China	5406593.9	
3	ESP	Spain	55082.2	
4	IDN	Indonesia	2407953.5	
5	IND	India	4580248.4	
6	IRN	Iran (Islamic Republic of)	116486.7	
7	ITA	Italy	114574.8	
8	JPN	Japan	330353.1	
9	KHM	Cambodia	436826.0	
10	KOR	Korea (the Republic of)	167862.2	
11	LAO	Lao People's Democratic Republic (the)	94826.8	
12	LKA	Sri Lanka	132640.0	
13	MMR	Myanmar	1059409.6	
14	MYS	Malaysia	121942.6	
15	NPL	Nepal	149262.2	
16	PAK	Pakistan	383529.3	
17	PHL	Philippines (the)	1557810.6	
18	PRK	Korea (the Democratic People's Republic of)	82823.3	
19	THA	Thailand	1554254.0	
20	TWN	Taiwan (Province of China)	45838.7	
21	USA	United States of America (the)	364728.0	
22	VNM	Viet Nam	1381744.4	
mean	NaN	mean	948478.0	

	2016	2017	2018	2019	\
0	1.093480e+06	1.154531e+06	1.144591e+06	1.144745e+06	
1	1.262782e+05	1.303229e+05	1.216152e+05	1.110848e+05	
2	5.399920e+06	5.400129e+06	5.302173e+06	5.214455e+06	
3	5.507310e+04	5.423240e+04	5.292500e+04	5.209850e+04	
4	2.387656e+06	2.425291e+06	2.405614e+06	2.257604e+06	
5	4.559136e+06	4.620791e+06	4.661155e+06	4.621417e+06	
6	1.310085e+05	8.723360e+04	9.393660e+04	9.610340e+04	
7	1.180030e+05	1.180030e+05	1.094638e+05	1.108951e+05	
8	3.264030e+05	3.237003e+05	3.222450e+05	3.205818e+05	
9	4.590031e+05	4.737453e+05	4.793627e+05	4.683789e+05	
10	1.635341e+05	1.584897e+05	1.549113e+05	1.532609e+05	
11	9.563000e+04	9.394070e+04	8.333360e+04	7.700550e+04	
12	1.217563e+05	8.445630e+04	1.110490e+05	1.021563e+05	
13	1.052288e+06	1.087030e+06	1.118850e+06	1.083100e+06	
14	1.232328e+05	1.226563e+05	1.252385e+05	1.224538e+05	
15	1.427237e+05	1.625746e+05	1.538908e+05	1.562154e+05	
16	3.813618e+05	4.060833e+05	3.934042e+05	4.247551e+05	
17	1.524292e+06	1.609862e+06	1.606048e+06	1.556226e+06	
18	8.344230e+04	8.459620e+04	8.394330e+04	8.293700e+04	

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19	1.703328e+06	1.714466e+06	1.702989e+06	1.553836e+06	
20	4.983830e+04	4.999120e+04	4.941410e+04	4.915200e+04	
21	4.386620e+05	3.362555e+05	4.121775e+05	3.501365e+05	
22	1.365174e+06	1.360552e+06	1.336231e+06	1.318431e+06	
mean	9.522272e+05	9.590840e+05	9.575896e+05	9.316100e+05	
	tCO2_2015_FAOSTAT	tCO2_2016_FAOSTAT	tCO2_2017_FAOSTAT	...	\
0	28282335.0	2.733701e+07	2.886328e+07	...	
1	3472757.5	3.156955e+06	3.258072e+06	...	
2	135164847.5	1.349980e+08	1.350032e+08	...	
3	1377055.0	1.376828e+06	1.355810e+06	...	
4	60198837.5	5.969141e+07	6.063226e+07	...	
5	114506210.0	1.139784e+08	1.155198e+08	...	
6	2912167.5	3.275212e+06	2.180840e+06	...	
7	2864370.0	2.950075e+06	2.950075e+06	...	
8	8258827.5	8.160075e+06	8.092508e+06	...	
9	10920650.0	1.147508e+07	1.184363e+07	...	
10	4196555.0	4.088352e+06	3.962243e+06	...	
11	2370670.0	2.390750e+06	2.348518e+06	...	
12	3316000.0	3.043908e+06	2.111408e+06	...	
13	26485240.0	2.630719e+07	2.717574e+07	...	
14	3048565.0	3.080820e+06	3.066408e+06	...	
15	3731555.0	3.568093e+06	4.064365e+06	...	
16	9588232.5	9.534045e+06	1.015208e+07	...	
17	38945265.0	3.810731e+07	4.024656e+07	...	
18	2070582.5	2.086058e+06	2.114905e+06	...	
19	38856350.0	4.258319e+07	4.286164e+07	...	
20	1145967.5	1.245958e+06	1.249780e+06	...	
21	9118200.0	1.096655e+07	8.406388e+06	...	
22	34543610.0	3.412934e+07	3.401379e+07	...	
mean	23711950.0	2.380568e+07	2.397710e+07	...	
	CO2_abs_percent_diff_2019	CO2_abs_percent_diff_means			\
0	57.606797	63.425917			
1	99.141295	93.222402			
2	7.189945	9.421813			
3	127.757634	124.705066			
4	56.234955	70.153790			
5	47.515394	32.086209			
6	1.152675	13.260902			
7	83.321649	80.187551			
8	31.555853	36.874729			
9	23.769728	14.304565			
10	14.321600	15.813819			
11	136.209544	145.627378			
12	18.612972	27.569003			
13	14.853751	14.697380			
14	14.753544	13.043224			
15	29.190154	50.437328			
16	42.334367	31.154530			
17	125.114417	121.748165			
18	15.169422	17.492669			
19	17.404685	22.551331			
20	58.617135	52.458032			
21	69.713577	78.850049			
22	3.761740	0.902428			

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mean	47.621862	49.129925	
	CO2_abs_percent_diff_totals	CO2_relative_percent_diff_2015	\
0	63.425917	-107.233570	
1	93.222402	-145.498958	
2	9.421813	-13.447527	
3	124.705066	79.277072	
4	70.153790	46.691296	
5	32.086209	-35.798018	
6	13.260902	24.674600	
7	80.187551	56.395576	
8	36.874729	30.212099	
9	14.304565	-13.424988	
10	15.813819	13.507737	
11	145.627378	82.482067	
12	27.569003	37.382189	
13	14.697380	-6.859731	
14	13.043224	13.287137	
15	50.437328	32.502761	
16	31.154530	-26.520489	
17	121.748165	77.968948	
18	17.492669	-38.030793	
19	22.551331	10.323662	
20	52.458032	-71.622589	
21	78.850049	55.821215	
22	0.902428	2.585962	
mean	49.129925	4.551116	
	CO2_relative_percent_diff_2016	CO2_relative_percent_diff_2017	\
0	-108.340089	-81.801753	
1	-145.821472	-185.841820	
2	-8.511442	-17.683689	
3	75.763074	77.554025	
4	57.149216	60.353916	
5	-16.456550	-34.791882	
6	29.927285	-10.281098	
7	58.155426	53.868303	
8	30.021068	16.313490	
9	-24.872761	4.652461	
10	22.059230	7.677136	
11	82.260369	87.565950	
12	16.904226	30.001182	
13	-22.666619	-10.868129	
14	9.922595	9.398001	
15	53.286238	50.291743	
16	-55.912882	-32.303912	
17	73.275770	76.166725	
18	-9.988016	-28.310326	
19	-4.559627	32.066348	
20	-62.307893	-74.143329	
21	63.101978	49.895259	
22	-8.687738	-3.372544	
mean	4.508756	3.322003	
	CO2_relative_percent_diff_2018	CO2_relative_percent_diff_2019	\
0	-87.073857	-80.912284	

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1	-205.638590	-196.594424
2	-2.108366	-7.458060
3	73.445258	77.958601
4	51.073546	43.893273
5	-41.376939	-62.321563
6	5.513463	1.146070
7	59.165480	58.817707
8	51.954183	27.255501
9	-16.667610	-26.975760
10	16.497815	13.364589
11	87.883944	81.026578
12	18.784377	17.028241
13	-22.665852	-16.045425
14	14.840462	13.739977
15	40.212459	25.472433
16	-15.207139	-53.701452
17	74.003201	76.966391
18	-3.195586	-16.414407
19	46.173612	16.011325
20	-64.692890	-82.919716
21	59.792699	51.694526
22	1.405162	3.692293
mean	6.179080	-1.533721

	CO2_relative_percent_diff_means	CO2_relative_percent_diff_totals
0	-92.881337	-92.881337
1	-174.610414	-174.610414
2	-9.887609	-9.887609
3	76.811284	76.811284
4	51.936188	51.936188
5	-38.217479	-38.217479
6	12.436318	12.436318
7	57.238482	57.238482
8	31.134372	31.134372
9	-15.406481	-15.406481
10	14.655057	14.655057
11	84.268427	84.268427
12	24.229137	24.229137
13	-15.863111	-15.863111
14	12.244674	12.244674
15	40.279401	40.279401
16	-36.903009	-36.903009
17	75.679166	75.679166
18	-19.169278	-19.169278
19	20.266184	20.266184
20	-71.109302	-71.109302
21	56.553728	56.553728
22	-0.906518	-0.906518
mean	3.599038	3.599038

[24 rows x 73 columns]

## 5.7.8 Recalculate Totals

```
merged_df.loc['total'] = merged_df[merged_df['country_name_FAOSTAT'] != 'mean'].
↳select_dtypes(np.number).sum()
```

```
merged_df.at['total','country_name_FAOSTAT'] = 'total'
merged_df.at['total','country_name_TRACE'] = 'total'
```

```
merged_df.reset_index(inplace=True, drop = True)
```

```
merged_df
```

	iso3_country	country_name_FAOSTAT	2015	\
0	BGD	Bangladesh	1131293.4	
1	BRA	Brazil	138910.3	
2	CHN	China	5406593.9	
3	ESP	Spain	55082.2	
4	IDN	Indonesia	2407953.5	
5	IND	India	4580248.4	
6	IRN	Iran (Islamic Republic of)	116486.7	
7	ITA	Italy	114574.8	
8	JPN	Japan	330353.1	
9	KHM	Cambodia	436826.0	
10	KOR	Korea (the Republic of)	167862.2	
11	LAO	Lao People's Democratic Republic (the)	94826.8	
12	LKA	Sri Lanka	132640.0	
13	MMR	Myanmar	1059409.6	
14	MYS	Malaysia	121942.6	
15	NPL	Nepal	149262.2	
16	PAK	Pakistan	383529.3	
17	PHL	Philippines (the)	1557810.6	
18	PRK	Korea (the Democratic People's Republic of)	82823.3	
19	THA	Thailand	1554254.0	
20	TWN	Taiwan (Province of China)	45838.7	
21	USA	United States of America (the)	364728.0	
22	VNM	Viet Nam	1381744.4	
23	NaN	mean	948478.0	
24	NaN	total	21814994.0	

	2016	2017	2018	2019	tCO2_2015_FAOSTAT	\
0	1.093480e+06	1.154531e+06	1.144591e+06	1.144745e+06	28282335.0	
1	1.262782e+05	1.303229e+05	1.216152e+05	1.110848e+05	3472757.5	
2	5.399920e+06	5.400129e+06	5.302173e+06	5.214455e+06	135164847.5	
3	5.507310e+04	5.423240e+04	5.292500e+04	5.209850e+04	1377055.0	
4	2.387656e+06	2.425291e+06	2.405614e+06	2.257604e+06	60198837.5	
5	4.559136e+06	4.620791e+06	4.661155e+06	4.621417e+06	114506210.0	
6	1.310085e+05	8.723360e+04	9.393660e+04	9.610340e+04	2912167.5	
7	1.180030e+05	1.180030e+05	1.094638e+05	1.108951e+05	2864370.0	
8	3.264030e+05	3.237003e+05	3.222450e+05	3.205818e+05	8258827.5	
9	4.590031e+05	4.737453e+05	4.793627e+05	4.683789e+05	10920650.0	
10	1.635341e+05	1.584897e+05	1.549113e+05	1.532609e+05	4196555.0	
11	9.563000e+04	9.394070e+04	8.333360e+04	7.700550e+04	2370670.0	
12	1.217563e+05	8.445630e+04	1.110490e+05	1.021563e+05	3316000.0	
13	1.052288e+06	1.087030e+06	1.118850e+06	1.083100e+06	26485240.0	

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14	1.232328e+05	1.226563e+05	1.252385e+05	1.224538e+05	3048565.0
15	1.427237e+05	1.625746e+05	1.538908e+05	1.562154e+05	3731555.0
16	3.813618e+05	4.060833e+05	3.934042e+05	4.247551e+05	9588232.5
17	1.524292e+06	1.609862e+06	1.606048e+06	1.556226e+06	38945265.0
18	8.344230e+04	8.459620e+04	8.394330e+04	8.293700e+04	2070582.5
19	1.703328e+06	1.714466e+06	1.702989e+06	1.553836e+06	38856350.0
20	4.983830e+04	4.999120e+04	4.941410e+04	4.915200e+04	1145967.5
21	4.386620e+05	3.362555e+05	4.121775e+05	3.501365e+05	9118200.0
22	1.365174e+06	1.360552e+06	1.336231e+06	1.318431e+06	34543610.0
23	9.522272e+05	9.590840e+05	9.575896e+05	9.316100e+05	23711950.0
24	2.190123e+07	2.205893e+07	2.202456e+07	2.142703e+07	545374850.0

	tCO2_2016_FAOSTAT	tCO2_2017_FAOSTAT	...	CO2_abs_percent_diff_2019	\
0	2.733701e+07	2.886328e+07	...	57.606797	
1	3.156955e+06	3.258072e+06	...	99.141295	
2	1.349980e+08	1.350032e+08	...	7.189945	
3	1.376828e+06	1.355810e+06	...	127.757634	
4	5.969141e+07	6.063226e+07	...	56.234955	
5	1.139784e+08	1.155198e+08	...	47.515394	
6	3.275212e+06	2.180840e+06	...	1.152675	
7	2.950075e+06	2.950075e+06	...	83.321649	
8	8.160075e+06	8.092508e+06	...	31.555853	
9	1.147508e+07	1.184363e+07	...	23.769728	
10	4.088352e+06	3.962243e+06	...	14.321600	
11	2.390750e+06	2.348518e+06	...	136.209544	
12	3.043908e+06	2.111408e+06	...	18.612972	
13	2.630719e+07	2.717574e+07	...	14.853751	
14	3.080820e+06	3.066408e+06	...	14.753544	
15	3.568093e+06	4.064365e+06	...	29.190154	
16	9.534045e+06	1.015208e+07	...	42.334367	
17	3.810731e+07	4.024656e+07	...	125.114417	
18	2.086058e+06	2.114905e+06	...	15.169422	
19	4.258319e+07	4.286164e+07	...	17.404685	
20	1.245958e+06	1.249780e+06	...	58.617135	
21	1.096655e+07	8.406388e+06	...	69.713577	
22	3.412934e+07	3.401379e+07	...	3.761740	
23	2.380568e+07	2.397710e+07	...	47.621862	
24	5.475306e+08	5.514733e+08	...	1095.302835	

	CO2_abs_percent_diff_means	CO2_abs_percent_diff_totals	\
0	63.425917	63.425917	
1	93.222402	93.222402	
2	9.421813	9.421813	
3	124.705066	124.705066	
4	70.153790	70.153790	
5	32.086209	32.086209	
6	13.260902	13.260902	
7	80.187551	80.187551	
8	36.874729	36.874729	
9	14.304565	14.304565	
10	15.813819	15.813819	
11	145.627378	145.627378	
12	27.569003	27.569003	
13	14.697380	14.697380	
14	13.043224	13.043224	
15	50.437328	50.437328	

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16	31.154530	31.154530
17	121.748165	121.748165
18	17.492669	17.492669
19	22.551331	22.551331
20	52.458032	52.458032
21	78.850049	78.850049
22	0.902428	0.902428
23	49.129925	49.129925
24	1129.988279	1129.988279
	CO2_relative_percent_diff_2015	CO2_relative_percent_diff_2016 \
0	-107.233570	-108.340089
1	-145.498958	-145.821472
2	-13.447527	-8.511442
3	79.277072	75.763074
4	46.691296	57.149216
5	-35.798018	-16.456550
6	24.674600	29.927285
7	56.395576	58.155426
8	30.212099	30.021068
9	-13.424988	-24.872761
10	13.507737	22.059230
11	82.482067	82.260369
12	37.382189	16.904226
13	-6.859731	-22.666619
14	13.287137	9.922595
15	32.502761	53.286238
16	-26.520489	-55.912882
17	77.968948	73.275770
18	-38.030793	-9.988016
19	10.323662	-4.559627
20	-71.622589	-62.307893
21	55.821215	63.101978
22	2.585962	-8.687738
23	4.551116	4.508756
24	104.675658	103.701389
	CO2_relative_percent_diff_2017	CO2_relative_percent_diff_2018 \
0	-81.801753	-87.073857
1	-185.841820	-205.638590
2	-17.683689	-2.108366
3	77.554025	73.445258
4	60.353916	51.073546
5	-34.791882	-41.376939
6	-10.281098	5.513463
7	53.868303	59.165480
8	16.313490	51.954183
9	4.652461	-16.667610
10	7.677136	16.497815
11	87.565950	87.883944
12	30.001182	18.784377
13	-10.868129	-22.665852
14	9.398001	14.840462
15	50.291743	40.212459
16	-32.303912	-15.207139
17	76.166725	74.003201

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18	-28.310326	-3.195586
19	32.066348	46.173612
20	-74.143329	-64.692890
21	49.895259	59.792699
22	-3.372544	1.405162
23	3.322003	6.179080
24	76.406059	142.118831
CO2_relative_percent_diff_2019 CO2_relative_percent_diff_means \		
0	-80.912284	-92.881337
1	-196.594424	-174.610414
2	-7.458060	-9.887609
3	77.958601	76.811284
4	43.893273	51.936188
5	-62.321563	-38.217479
6	1.146070	12.436318
7	58.817707	57.238482
8	27.255501	31.134372
9	-26.975760	-15.406481
10	13.364589	14.655057
11	81.026578	84.268427
12	17.028241	24.229137
13	-16.045425	-15.863111
14	13.739977	12.244674
15	25.472433	40.279401
16	-53.701452	-36.903009
17	76.966391	75.679166
18	-16.414407	-19.169278
19	16.011325	20.266184
20	-82.919716	-71.109302
21	51.694526	56.553728
22	3.692293	-0.906518
23	-1.533721	3.599038
24	-35.275585	82.777881
CO2_relative_percent_diff_totals		
0	-92.881337	
1	-174.610414	
2	-9.887609	
3	76.811284	
4	51.936188	
5	-38.217479	
6	12.436318	
7	57.238482	
8	31.134372	
9	-15.406481	
10	14.655057	
11	84.268427	
12	24.229137	
13	-15.863111	
14	12.244674	
15	40.279401	
16	-36.903009	
17	75.679166	
18	-19.169278	
19	20.266184	

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```

20          -71.109302
21          56.553728
22          -0.906518
23           3.599038
24          82.777881

```

```
[25 rows x 73 columns]
```

### 5.7.9 Merged Data to File

```

outfile = "/Users/jnapolitano/Projects/wattime-takehome/data/MERGED_DATA.csv"

merged_df.to_csv(outfile)

```

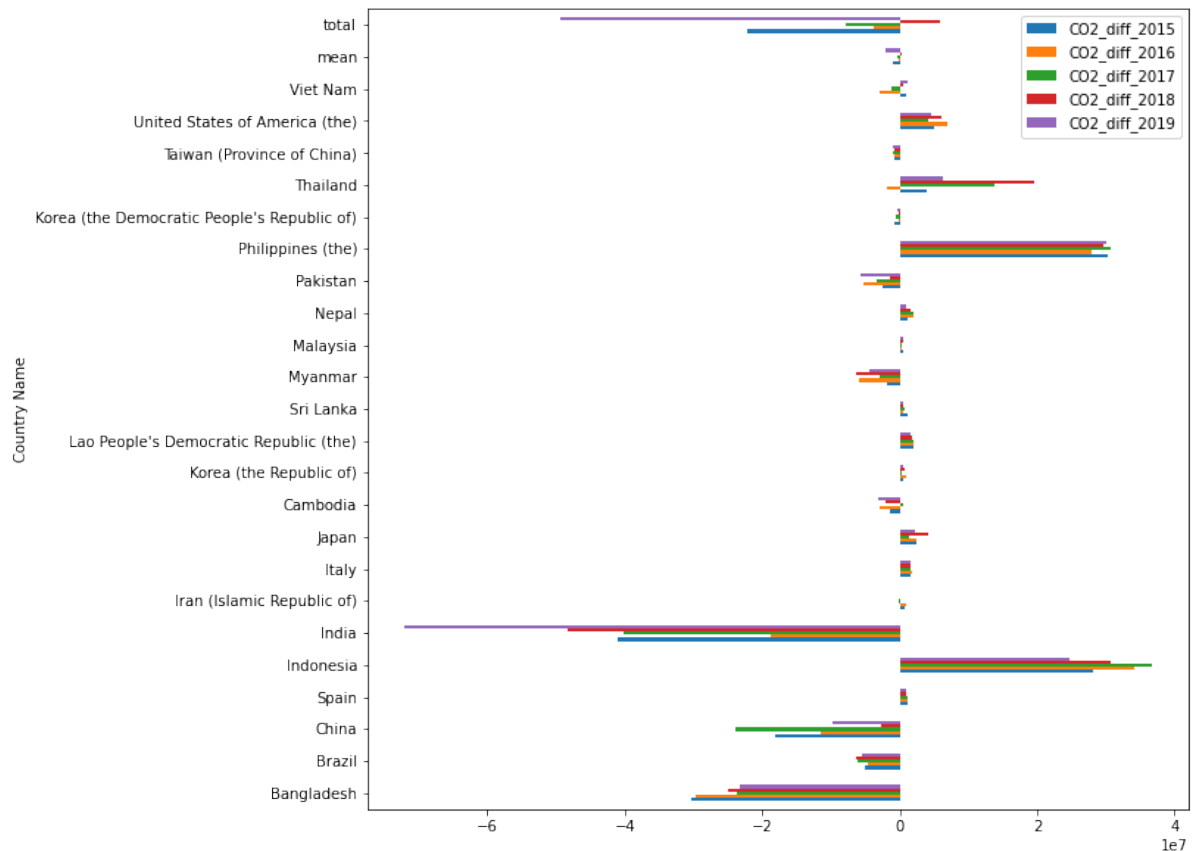
### 5.7.10 CO2 Difference Plots

```

merged_df.plot(kind = "barh", x = 'country_name_FAOSTAT', y = ["CO2_diff_2015", "CO2_
diff_2016", "CO2_diff_2017", "CO2_diff_2018", "CO2_diff_2019"], xlabel =
"Country Name", ylabel = "Tonnes CH4", figsize = (10,10))

```

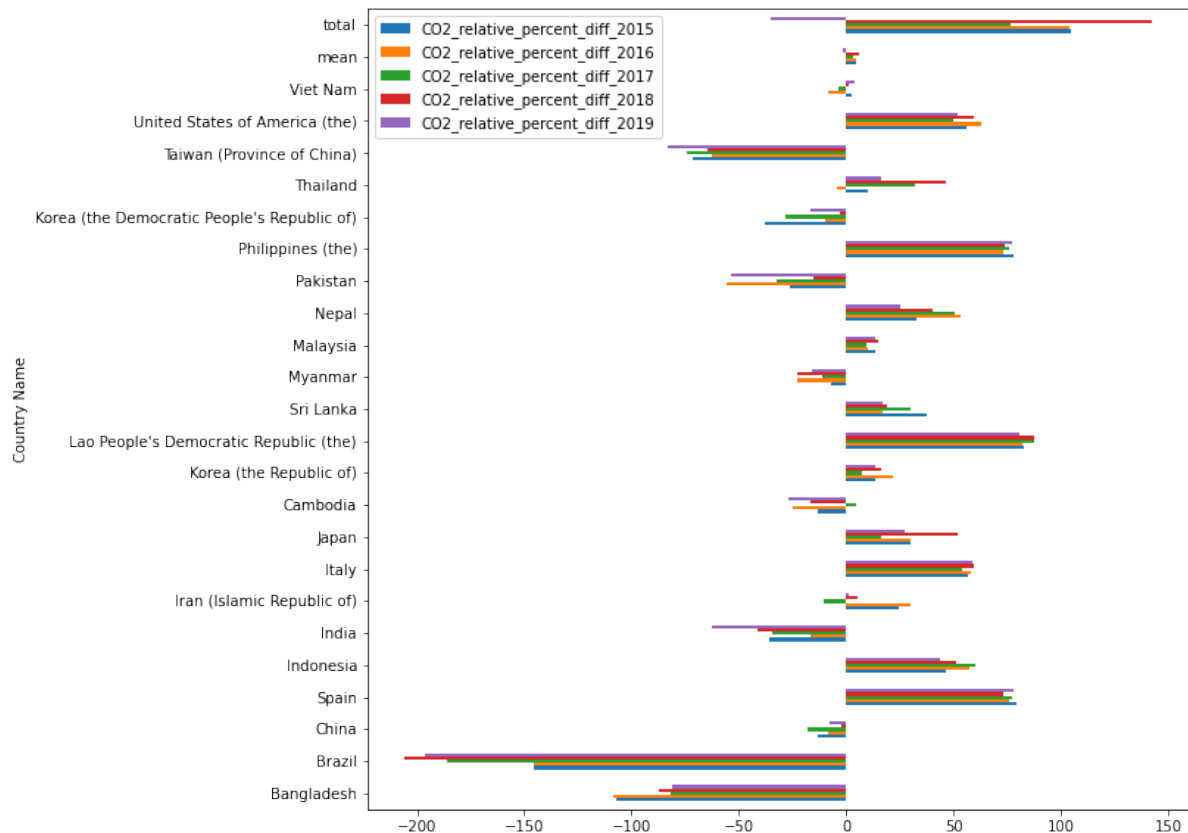
```
<AxesSubplot:ylabel='Country Name'>
```



### 5.7.11 Percent Difference Plot

```
merged_df.plot(kind = "barh", x = 'country_name_FAOSTAT', y = ["CO2_relative_percent_diff_2015", "CO2_relative_percent_diff_2016", "CO2_relative_percent_diff_2017", "CO2_relative_percent_diff_2018", "CO2_relative_percent_diff_2019"], xlabel = "Country Name", ylabel = "Tonnes CH4", figsize=(10,10))
```

```
<AxesSubplot:ylabel='Country Name'>
```



## 5.8 Impressions

### 5.8.1 Initial

The percent difference and the tonnage difference do not support each other. I need to recalculate the totals section to ensure that we are doing things correctly.

I need to confirm the values, but I'm initially impressed by the fact that the faostat data reports higher values than the malaysia data on average. According to the included paper this should not be the case.

The quote included below states the problems with the malaysia methodology.

“The difference between harvested rice cultivation area from statistical data and remote-sensing estimates can be due to two factors: (i) MODIS data which have moderate spatial resolution lead to mixed pixels, where rice fields and non-rice fields are combined. This can overestimate area, especially in lowland regions and have a low ability to detect small rice field patches in upland regions (Frolking et al 1999, Seto et al 2000); and (ii) political and policy factors (Yan et al., 2019)

such as determination of the amount of subsidies for fertilizers and evaluation of achievement of government programs in the agricultural sector. Other factors that contribute to discrepancy in CH<sub>4</sub> emission are from different emission and scale factors that are related to water regime and organic amendment. These values give high uncertainty since the availability of these data are limited and quite variable.”

### Verified

I calculated differences totals and means per dataframe to ensure accuracy prior to the join. I also dropped precalculated values when joining to ensure that the aggregation algorithms to not modify the results.

The data is now consistent and supports the findings of the University of Malaysia Paper

## MERGING TRACE AND FAOSTAT DATA WITH GEOGRAPHIC DATA FOR MAPPING

```
import pandas as pd
import matplotlib.pyplot as plt
import geopandas as gpd
import folium
import contextily as cx
from shapely.geometry import Point, LineString, Polygon
import numpy as np
from scipy.spatial import cKDTree
from geopy.distance import distance
import scipy.stats as stats
```

```
/Users/jnapolitano/venvs/finance/lib/python3.9/site-packages/geopandas/_compat.
py:111: UserWarning: The Shapely GEOS version (3.10.2-CAPI-1.16.0) is
incompatible with the GEOS version PyGEOS was compiled with (3.10.1-CAPI-1.16.0).
Conversions between both will be slow.
warnings.warn(
```

### 6.1 Shape Data

```
gisfilepath = "/Users/jnapolitano/Projects/wattime-takehome/data/country_shapefiles/
World_Countries__Generalized_.shp"
```

```
countries_df = gpd.read_file(gisfilepath)
countries_df = countries_df.to_crs(epsg=3857)
```

```
world = gpd.read_file(gpd.datasets.get_path('naturalearth_lowres'))
cities = gpd.read_file(gpd.datasets.get_path('naturalearth_cities'))
```

```
world.rename({"iso_a3" : 'iso3_country'}, axis =1, inplace=True)
```

```
world
```

```

      pop_est      continent      name iso3_country \
0      920938      Oceania      Fiji      FJI
1      53950935      Africa      Tanzania      TZA
2      603253      Africa      W. Sahara      ESH
3      35623680      North America      Canada      CAN
4      326625791      North America      United States of America      USA
..      ...      ...      ...      ...
172      7111024      Europe      Serbia      SRB
173      642550      Europe      Montenegro      MNE
174      1895250      Europe      Kosovo      -99
175      1218208      North America      Trinidad and Tobago      TTO
176      13026129      Africa      S. Sudan      SSD

      gdp_md_est      geometry
0      8374.0      MULTIPOLYGON (((180.00000 -16.06713, 180.00000...
1      150600.0      POLYGON ((33.90371 -0.95000, 34.07262 -1.05982...
2      906.5      POLYGON ((-8.66559 27.65643, -8.66512 27.58948...
3      1674000.0      MULTIPOLYGON (((-122.84000 49.00000, -122.9742...
4      18560000.0      MULTIPOLYGON (((-122.84000 49.00000, -120.0000...
..      ...      ...
172      101800.0      POLYGON ((18.82982 45.90887, 18.82984 45.90888...
173      10610.0      POLYGON ((20.07070 42.58863, 19.80161 42.50009...
174      18490.0      POLYGON ((20.59025 41.85541, 20.52295 42.21787...
175      43570.0      POLYGON ((-61.68000 10.76000, -61.10500 10.890...
176      20880.0      POLYGON ((30.83385 3.50917, 29.95350 4.17370, ...

[177 rows x 6 columns]

```

## 6.2 FAOSTAT DATA

```

filepath = "/Users/jnapolitano/Projects/watttime-takehome/data/FAOSTAT_DATA.csv"

faostat_emissions_df = pd.read_csv(filepath)
faostat_emissions_df.drop('Unnamed: 0', axis = 1, inplace = True)
faostat_emissions_df = faostat_emissions_df[(faostat_emissions_df["country_name"] !=
↳ 'Total') & (faostat_emissions_df["country_name"] != "mean")]
faostat_emissions_df.dropna(how="any", axis=1, inplace= True)

```

```
faostat_emissions_df
```

```

      iso3_country      country_name      2015 \
0      BGD      Bangladesh      1131293.4
1      BRA      Brazil      138910.3
2      CHN      China      5406593.9
3      ESP      Spain      55082.2
4      IDN      Indonesia      2407953.5
5      IND      India      4580248.4
6      IRN      Iran (Islamic Republic of)      116486.7
7      ITA      Italy      114574.8
8      JPN      Japan      330353.1
9      KHM      Cambodia      436826.0
10     KOR      Korea (the Republic of)      167862.2

```

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11	LAO	Lao People's Democratic Republic (the)	94826.8
12	LKA	Sri Lanka	132640.0
13	MMR	Myanmar	1059409.6
14	MYS	Malaysia	121942.6
15	NPL	Nepal	149262.2
16	PAK	Pakistan	383529.3
17	PHL	Philippines (the)	1557810.6
18	PRK	Korea (the Democratic People's Republic of)	82823.3
19	THA	Thailand	1554254.0
20	TWN	Taiwan (Province of China)	45838.7
21	USA	United States of America (the)	364728.0
22	VNM	Viet Nam	1381744.4

	2016	2017	2018	2019	tCO2_2015	tCO2_2016	\
0	1093480.4	1154531.0	1144591.0	1144745.4	28282335.0	27337010.0	
1	126278.2	130322.9	121615.2	111084.8	3472757.5	3156955.0	
2	5399920.0	5400129.0	5302173.1	5214454.7	135164847.5	134998000.0	
3	55073.1	54232.4	52925.0	52098.5	1377055.0	1376827.5	
4	2387656.4	2425290.6	2405613.8	2257604.3	60198837.5	59691410.0	
5	4559136.4	4620790.8	4661154.9	4621416.8	114506210.0	113978410.0	
6	131008.5	87233.6	93936.6	96103.4	2912167.5	3275212.5	
7	118003.0	118003.0	109463.8	110895.1	2864370.0	2950075.0	
8	326403.0	323700.3	322245.0	320581.8	8258827.5	8160075.0	
9	459003.1	473745.3	479362.7	468378.9	10920650.0	11475077.5	
10	163534.1	158489.7	154911.3	153260.9	4196555.0	4088352.5	
11	95630.0	93940.7	83333.6	77005.5	2370670.0	2390750.0	
12	121756.3	84456.3	111049.0	102156.3	3316000.0	3043907.5	
13	1052287.7	1087029.5	1118850.0	1083100.3	26485240.0	26307192.5	
14	123232.8	122656.3	125238.5	122453.8	3048565.0	3080820.0	
15	142723.7	162574.6	153890.8	156215.4	3731555.0	3568092.5	
16	381361.8	406083.3	393404.2	424755.1	9588232.5	9534045.0	
17	1524292.5	1609862.5	1606047.8	1556225.8	38945265.0	38107312.5	
18	83442.3	84596.2	83943.3	82937.0	2070582.5	2086057.5	
19	1703327.7	1714465.6	1702989.1	1553835.5	38856350.0	42583192.5	
20	49838.3	49991.2	49414.1	49152.0	1145967.5	1245957.5	
21	438662.0	336255.5	412177.5	350136.5	9118200.0	10966550.0	
22	1365173.8	1360551.6	1336231.2	1318431.1	34543610.0	34129345.0	

	tCO2_2017	tCO2_2018	tCO2_2019	Mean_CH4	Total_CH4	\
0	28863275.0	28614775.0	28618635.0	1133728.24	5668641.2	
1	3258072.5	3040380.0	2777120.0	125642.28	628211.4	
2	135003225.0	132554327.5	130361367.5	5344654.14	26723270.7	
3	1355810.0	1323125.0	1302462.5	53882.24	269411.2	
4	60632265.0	60140345.0	56440107.5	2376823.72	11884118.6	
5	115519770.0	116528872.5	115535420.0	4608549.46	23042747.3	
6	2180840.0	2348415.0	2402585.0	104953.76	524768.8	
7	2950075.0	2736595.0	2772377.5	114187.94	570939.7	
8	8092507.5	8056125.0	8014545.0	324656.64	1623283.2	
9	11843632.5	11984067.5	11709472.5	463463.20	2317316.0	
10	3962242.5	3872782.5	3831522.5	159611.64	798058.2	
11	2348517.5	2083340.0	1925137.5	88947.32	444736.6	
12	2111407.5	2776225.0	2553907.5	110411.58	552057.9	
13	27175737.5	27971250.0	27077507.5	1080135.42	5400677.1	
14	3066407.5	3130962.5	3061345.0	123104.80	615524.0	
15	4064365.0	3847270.0	3905385.0	152933.34	764666.7	
16	10152082.5	9835105.0	10618877.5	397826.74	1989133.7	

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17	40246562.5	40151195.0	38905645.0	1570847.84	7854239.2
18	2114905.0	2098582.5	2073425.0	83548.42	417742.1
19	42861640.0	42574727.5	38845887.5	1645774.38	8228871.9
20	1249780.0	1235352.5	1228800.0	48846.86	244234.3
21	8406387.5	10304437.5	8753412.5	380391.90	1901959.5
22	34013790.0	33405780.0	32960777.5	1352426.42	6762132.1

	Mean_CO2	Total_CO2
0	28343206.0	141716030.0
1	3141057.0	15705285.0
2	133616353.5	668081767.5
3	1347056.0	6735280.0
4	59420593.0	297102965.0
5	115213736.5	576068682.5
6	2623844.0	13119220.0
7	2854698.5	14273492.5
8	8116416.0	40582080.0
9	11586580.0	57932900.0
10	3990291.0	19951455.0
11	2223683.0	11118415.0
12	2760289.5	13801447.5
13	27003385.5	135016927.5
14	3077620.0	15388100.0
15	3823333.5	19116667.5
16	9945668.5	49728342.5
17	39271196.0	196355980.0
18	2088710.5	10443552.5
19	41144359.5	205721797.5
20	1221171.5	6105857.5
21	9509797.5	47548987.5
22	33810660.5	169053302.5

## 6.2.1 FAOSTAT Country Merge

```
faostat_merged_df = world.merge(faostat_emissions_df, on='iso3_country', how='right',
➔sort=True)
```

```
faostat_merged_df
#trace_merged_df.dropna(axis=1, how='any')
```

	pop_est	continent	name	iso3_country	\
0	157826578	Asia	Bangladesh	BGD	
1	207353391	South America	Brazil	BRA	
2	1379302771	Asia	China	CHN	
3	48958159	Europe	Spain	ESP	
4	260580739	Asia	Indonesia	IDN	
5	1281935911	Asia	India	IND	
6	82021564	Asia	Iran	IRN	
7	62137802	Europe	Italy	ITA	
8	126451398	Asia	Japan	JPN	
9	16204486	Asia	Cambodia	KHM	
10	51181299	Asia	South Korea	KOR	

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11	7126706	Asia	Laos	LAO
12	22409381	Asia	Sri Lanka	LKA
13	55123814	Asia	Myanmar	MMR
14	31381992	Asia	Malaysia	MYS
15	29384297	Asia	Nepal	NPL
16	204924861	Asia	Pakistan	PAK
17	104256076	Asia	Philippines	PHL
18	25248140	Asia	North Korea	PRK
19	68414135	Asia	Thailand	THA
20	23508428	Asia	Taiwan	TWN
21	326625791	North America	United States of America	USA
22	96160163	Asia	Vietnam	VNM

	gdp_md_est	geometry	\
0	628400.0	POLYGON ((92.67272 22.04124, 92.65226 21.32405...	
1	3081000.0	POLYGON ((-53.37366 -33.76838, -53.65054 -33.2...	
2	21140000.0	MULTIPOLYGON (((109.47521 18.19770, 108.65521 ...	
3	1690000.0	POLYGON ((-7.45373 37.09779, -7.53711 37.42890...	
4	3028000.0	MULTIPOLYGON (((141.00021 -2.60015, 141.01706 ...	
5	8721000.0	POLYGON ((97.32711 28.26158, 97.40256 27.88254...	
6	1459000.0	POLYGON ((48.56797 29.92678, 48.01457 30.45246...	
7	2221000.0	MULTIPOLYGON (((10.44270 46.89355, 11.04856 46...	
8	4932000.0	MULTIPOLYGON (((141.88460 39.18086, 140.95949 ...	
9	58940.0	POLYGON ((102.58493 12.18659, 102.34810 13.394...	
10	1929000.0	POLYGON ((126.17476 37.74969, 126.23734 37.840...	
11	40960.0	POLYGON ((107.38273 14.20244, 106.49637 14.570...	
12	236700.0	POLYGON ((81.78796 7.52306, 81.63732 6.48178, ...	
13	311100.0	POLYGON ((100.11599 20.41785, 99.54331 20.1866...	
14	863000.0	MULTIPOLYGON (((100.08576 6.46449, 100.25960 6...	
15	71520.0	POLYGON ((88.12044 27.87654, 88.04313 27.44582...	
16	988200.0	POLYGON ((77.83745 35.49401, 76.87172 34.65354...	
17	801900.0	MULTIPOLYGON (((120.83390 12.70450, 120.32344 ...	
18	40000.0	MULTIPOLYGON (((130.78000 42.22001, 130.78000 ...	
19	1161000.0	POLYGON ((105.21878 14.27321, 104.28142 14.416...	
20	1127000.0	POLYGON ((121.77782 24.39427, 121.17563 22.790...	
21	18560000.0	MULTIPOLYGON (((-122.84000 49.00000, -120.0000...	
22	594900.0	POLYGON ((104.33433 10.48654, 105.19991 10.889...	

	country_name	2015	2016	\
0	Bangladesh	1131293.4	1093480.4	
1	Brazil	138910.3	126278.2	
2	China	5406593.9	5399920.0	
3	Spain	55082.2	55073.1	
4	Indonesia	2407953.5	2387656.4	
5	India	4580248.4	4559136.4	
6	Iran (Islamic Republic of)	116486.7	131008.5	
7	Italy	114574.8	118003.0	
8	Japan	330353.1	326403.0	
9	Cambodia	436826.0	459003.1	
10	Korea (the Republic of)	167862.2	163534.1	
11	Lao People's Democratic Republic (the)	94826.8	95630.0	
12	Sri Lanka	132640.0	121756.3	
13	Myanmar	1059409.6	1052287.7	
14	Malaysia	121942.6	123232.8	
15	Nepal	149262.2	142723.7	
16	Pakistan	383529.3	381361.8	

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17				Philippines (the)	1557810.6	1524292.5	
18	Korea (the Democratic People's Republic of)				82823.3	83442.3	
19				Thailand	1554254.0	1703327.7	
20				Taiwan (Province of China)	45838.7	49838.3	
21				United States of America (the)	364728.0	438662.0	
22				Viet Nam	1381744.4	1365173.8	
	2017	...	2019	tCO2_2015	tCO2_2016	tCO2_2017	\
0	1154531.0	...	1144745.4	28282335.0	27337010.0	28863275.0	
1	130322.9	...	111084.8	3472757.5	3156955.0	3258072.5	
2	5400129.0	...	5214454.7	135164847.5	134998000.0	135003225.0	
3	54232.4	...	52098.5	1377055.0	1376827.5	1355810.0	
4	2425290.6	...	2257604.3	60198837.5	59691410.0	60632265.0	
5	4620790.8	...	4621416.8	114506210.0	113978410.0	115519770.0	
6	87233.6	...	96103.4	2912167.5	3275212.5	2180840.0	
7	118003.0	...	110895.1	2864370.0	2950075.0	2950075.0	
8	323700.3	...	320581.8	8258827.5	8160075.0	8092507.5	
9	473745.3	...	468378.9	10920650.0	11475077.5	11843632.5	
10	158489.7	...	153260.9	4196555.0	4088352.5	3962242.5	
11	93940.7	...	77005.5	2370670.0	2390750.0	2348517.5	
12	84456.3	...	102156.3	3316000.0	3043907.5	2111407.5	
13	1087029.5	...	1083100.3	26485240.0	26307192.5	27175737.5	
14	122656.3	...	122453.8	3048565.0	3080820.0	3066407.5	
15	162574.6	...	156215.4	3731555.0	3568092.5	4064365.0	
16	406083.3	...	424755.1	9588232.5	9534045.0	10152082.5	
17	1609862.5	...	1556225.8	38945265.0	38107312.5	40246562.5	
18	84596.2	...	82937.0	2070582.5	2086057.5	2114905.0	
19	1714465.6	...	1553835.5	38856350.0	42583192.5	42861640.0	
20	49991.2	...	49152.0	1145967.5	1245957.5	1249780.0	
21	336255.5	...	350136.5	9118200.0	10966550.0	8406387.5	
22	1360551.6	...	1318431.1	34543610.0	34129345.0	34013790.0	
	tCO2_2018		tCO2_2019	Mean_CH4	Total_CH4	Mean_CO2	Total_CO2
0	28614775.0		28618635.0	1133728.24	5668641.2	28343206.0	141716030.0
1	3040380.0		2777120.0	125642.28	628211.4	3141057.0	15705285.0
2	132554327.5		130361367.5	5344654.14	26723270.7	133616353.5	668081767.5
3	1323125.0		1302462.5	53882.24	269411.2	1347056.0	6735280.0
4	60140345.0		56440107.5	2376823.72	11884118.6	59420593.0	297102965.0
5	116528872.5		115535420.0	4608549.46	23042747.3	115213736.5	576068682.5
6	2348415.0		2402585.0	104953.76	524768.8	2623844.0	13119220.0
7	2736595.0		2772377.5	114187.94	570939.7	2854698.5	14273492.5
8	8056125.0		8014545.0	324656.64	1623283.2	8116416.0	40582080.0
9	11984067.5		11709472.5	463463.20	2317316.0	11586580.0	57932900.0
10	3872782.5		3831522.5	159611.64	798058.2	3990291.0	19951455.0
11	2083340.0		1925137.5	88947.32	444736.6	2223683.0	11118415.0
12	2776225.0		2553907.5	110411.58	552057.9	2760289.5	13801447.5
13	27971250.0		27077507.5	1080135.42	5400677.1	27003385.5	135016927.5
14	3130962.5		3061345.0	123104.80	615524.0	3077620.0	15388100.0
15	3847270.0		3905385.0	152933.34	764666.7	3823333.5	19116667.5
16	9835105.0		10618877.5	397826.74	1989133.7	9945668.5	49728342.5
17	40151195.0		38905645.0	1570847.84	7854239.2	39271196.0	196355980.0
18	2098582.5		2073425.0	83548.42	417742.1	2088710.5	10443552.5
19	42574727.5		38845887.5	1645774.38	8228871.9	41144359.5	205721797.5
20	1235352.5		1228800.0	48846.86	244234.3	1221171.5	6105857.5
21	10304437.5		8753412.5	380391.90	1901959.5	9509797.5	47548987.5
22	33405780.0		32960777.5	1352426.42	6762132.1	33810660.5	169053302.5

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```
[23 rows x 21 columns]
```

```
faostat_merged_df = faostat_merged_df[faostat_merged_df['continent'] != 'Antarctica'].
↳copy()
```

## 6.2.2 FAOSTAT Merge to File

```
gisout = "/Users/jnapolitano/Projects/wattime-takehome/data/FAOSTAT_GEO.geojson"
faostat_merged_df.to_file(gisout, driver="GeoJSON")
```

```
/Users/jnapolitano/venvs/finance/lib/python3.9/site-packages/geopandas/io/file.
↳py:362: FutureWarning: pandas.Int64Index is deprecated and will be removed from
↳pandas in a future version. Use pandas.Index with the appropriate dtype instead.
pd.Int64Index,
```

## 6.3 TRACE Data

```
filepath = "/Users/jnapolitano/Projects/wattime-takehome/data/TRACE_DATA.csv"

trace_emissions_df = pd.read_csv(filepath)
trace_emissions_df.drop('Unnamed: 0', axis = 1, inplace = True)
trace_emissions_df = trace_emissions_df[(trace_emissions_df["country_name"] != 'Total
↳') & (trace_emissions_df["country_name"] != "mean")]
```

```
trace_emissions_df
```

	iso3_country	country_name	tCH4_2015 \
0	BGD	Bangladesh	2.344420e+06
1	BRA	Brazil	3.410233e+05
2	CHN	China	6.133647e+06
3	ESP	Spain	1.141464e+04
4	IDN	Indonesia	1.283649e+06
5	IND	India	6.219887e+06
6	IRN	Iran (Islamic Republic of)	8.774407e+04
7	ITA	Italy	4.995968e+04
8	JPN	Japan	2.305465e+05
9	KHM	Cambodia	4.954698e+05
10	KOR	Korea (the Republic of)	1.451878e+05
11	LAO	Lao People's Democratic Republic (the)	1.661169e+04
12	LKA	Sri Lanka	8.305626e+04
13	MMR	Myanmar	1.132082e+06
14	MYS	Malaysia	1.057399e+05
15	NPL	Nepal	1.007479e+05
16	PAK	Pakistan	4.852431e+05
17	PHL	Philippines (the)	3.432021e+05
18	PRK	Korea (the Democratic People's Republic of)	1.143217e+05

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19	THA	Thailand	1.393798e+06
20	TWN	Taiwan (Province of China)	7.866956e+04
21	USA	United States of America (the)	1.611324e+05
22	VNM	Viet Nam	1.346013e+06

	tCH4_2016	tCH4_2017	tCH4_2018	tCH4_2019	tCH4_2020 \
0	2.278158e+06	2.098958e+06	2.141231e+06	2.070985e+06	2.106781e+06
1	3.104189e+05	3.725173e+05	3.717030e+05	3.294713e+05	4.902874e+05
2	5.859531e+06	6.355071e+06	5.413962e+06	5.603352e+06	6.402353e+06
3	1.334803e+04	1.217299e+04	1.405410e+04	1.148324e+04	1.305461e+04
4	1.023129e+06	9.615327e+05	1.176982e+06	1.266668e+06	1.188195e+06
5	5.309413e+06	6.228451e+06	6.589798e+06	7.501556e+06	7.599764e+06
6	9.180121e+04	9.620217e+04	8.875744e+04	9.500199e+04	9.600254e+04
7	4.937785e+04	5.443679e+04	4.469902e+04	4.566914e+04	5.101547e+04
8	2.284133e+05	2.708935e+05	1.548252e+05	2.332056e+05	2.835167e+05
9	5.731698e+05	4.517045e+05	5.592610e+05	5.947277e+05	6.412802e+05
10	1.274597e+05	1.463222e+05	1.293543e+05	1.327782e+05	1.165467e+05
11	1.696441e+04	1.168063e+04	1.009675e+04	1.461058e+04	2.136270e+04
12	1.011743e+05	5.911841e+04	9.018914e+04	8.476088e+04	9.248238e+04
13	1.290806e+06	1.205169e+06	1.372447e+06	1.256888e+06	1.221904e+06
14	1.110049e+05	1.111291e+05	1.066525e+05	1.056287e+05	1.127141e+05
15	6.667161e+04	8.081300e+04	9.200752e+04	1.164235e+05	7.168401e+04
16	5.945922e+05	5.372641e+05	4.532297e+05	6.528548e+05	6.401201e+05
17	4.073554e+05	3.836830e+05	4.175210e+05	3.584550e+05	4.462836e+05
18	9.177653e+04	1.085457e+05	8.662578e+04	9.655062e+04	8.581038e+04
19	1.780993e+06	1.164699e+06	9.166575e+05	1.305046e+06	1.520788e+06
20	8.089149e+04	8.705634e+04	8.138151e+04	8.990870e+04	8.333327e+04
21	1.618576e+05	1.684799e+05	1.657254e+05	1.691351e+05	1.941455e+05
22	1.483777e+06	1.406437e+06	1.317455e+06	1.269751e+06	1.374450e+06

	tCH4_2021	tCO2_2015	tCO2_2016	tCO2_2017	tCO2_2018 \
0	1.983974e+06	5.861049e+07	5.695395e+07	5.247394e+07	5.353076e+07
1	4.544874e+05	8.525583e+06	7.760473e+06	9.312934e+06	9.292575e+06
2	6.068210e+06	1.533412e+08	1.464883e+08	1.588768e+08	1.353491e+08
3	8.531579e+03	2.853661e+05	3.337007e+05	3.043248e+05	3.513524e+05
4	1.009936e+06	3.209122e+07	2.557824e+07	2.403832e+07	2.942454e+07
5	6.567960e+06	1.554972e+08	1.327353e+08	1.557113e+08	1.647450e+08
6	9.053525e+04	2.193602e+06	2.295030e+06	2.405054e+06	2.218936e+06
7	5.089759e+04	1.248992e+06	1.234446e+06	1.360920e+06	1.117475e+06
8	1.574007e+05	5.763662e+06	5.710333e+06	6.772337e+06	3.870631e+06
9	5.644891e+05	1.238675e+07	1.432925e+07	1.129261e+07	1.398153e+07
10	1.013006e+05	3.629695e+06	3.186493e+06	3.658056e+06	3.233858e+06
11	1.475014e+04	4.152924e+05	4.241102e+05	2.920158e+05	2.524186e+05
12	8.466966e+04	2.076407e+06	2.529358e+06	1.477960e+06	2.254728e+06
13	1.289837e+06	2.830206e+07	3.227014e+07	3.012923e+07	3.431117e+07
14	1.069696e+05	2.643498e+06	2.775123e+06	2.778227e+06	2.666313e+06
15	4.811408e+04	2.518697e+06	1.666790e+06	2.020325e+06	2.300188e+06
16	4.849205e+05	1.213108e+07	1.486480e+07	1.343160e+07	1.133074e+07
17	4.383270e+05	8.580052e+06	1.018389e+07	9.592074e+06	1.043803e+07
18	7.735988e+04	2.858041e+06	2.294413e+06	2.713641e+06	2.165645e+06
19	8.528673e+05	3.484495e+07	4.452483e+07	2.911748e+07	2.291644e+07
20	6.619861e+04	1.966739e+06	2.022287e+06	2.176408e+06	2.034538e+06
21	1.634842e+05	4.028310e+06	4.046440e+06	4.211999e+06	4.143136e+06
22	1.502787e+06	3.365033e+07	3.709441e+07	3.516092e+07	3.293637e+07

tCO2_2019	Mean_CH4	Total_CH4	Mean_CO2	Total_CO2
-----------	----------	-----------	----------	-----------

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0	5.177463e+07	2.186750e+06	1.093375e+07	5.466875e+07	2.733438e+08
1	8.236783e+06	3.450268e+05	1.725134e+06	8.625670e+06	4.312835e+07
2	1.400838e+08	5.873113e+06	2.936556e+07	1.468278e+08	7.341391e+08
3	2.870810e+05	1.249460e+04	6.247300e+04	3.123650e+05	1.561825e+06
4	3.166670e+07	1.142392e+06	5.711960e+06	2.855980e+07	1.427990e+08
5	1.875389e+08	6.369821e+06	3.184910e+07	1.592455e+08	7.962276e+08
6	2.375050e+06	9.190138e+04	4.595069e+05	2.297534e+06	1.148767e+07
7	1.141729e+06	4.882850e+04	2.441425e+05	1.220712e+06	6.103562e+06
8	5.830141e+06	2.235768e+05	1.117884e+06	5.589421e+06	2.794710e+07
9	1.486819e+07	5.348666e+05	2.674333e+06	1.337166e+07	6.685832e+07
10	3.319455e+06	1.362205e+05	6.811023e+05	3.405512e+06	1.702756e+07
11	3.652645e+05	1.399281e+04	6.996406e+04	3.498203e+05	1.749102e+06
12	2.119022e+06	8.365981e+04	4.182990e+05	2.091495e+06	1.045748e+07
13	3.142221e+07	1.251478e+06	6.257392e+06	3.128696e+07	1.564348e+08
14	2.640717e+06	1.080310e+05	5.401551e+05	2.700775e+06	1.350388e+07
15	2.910588e+06	9.133271e+04	4.566635e+05	2.283318e+06	1.141659e+07
16	1.632137e+07	5.446368e+05	2.723184e+06	1.361592e+07	6.807960e+07
17	8.961374e+06	3.820433e+05	1.910216e+06	9.551082e+06	4.775541e+07
18	2.413765e+06	9.956405e+04	4.978202e+05	2.489101e+06	1.244551e+07
19	3.262615e+07	1.312239e+06	6.561194e+06	3.280597e+07	1.640298e+08
20	2.247717e+06	8.358152e+04	4.179076e+05	2.089538e+06	1.044769e+07
21	4.228377e+06	1.652661e+05	8.263305e+05	4.131652e+06	2.065826e+07
22	3.174377e+07	1.364686e+06	6.823432e+06	3.411716e+07	1.705858e+08

### 6.3.1 TRACE Country Merge

```
trace_merged_df = world.merge(trace_emissions_df, on='iso3_country', how='right',
↪sort=True)
```

```
trace_merged_df
```

	pop_est	continent	name	iso3_country	\
0	157826578	Asia	Bangladesh	BGD	
1	207353391	South America	Brazil	BRA	
2	1379302771	Asia	China	CHN	
3	48958159	Europe	Spain	ESP	
4	260580739	Asia	Indonesia	IDN	
5	1281935911	Asia	India	IND	
6	82021564	Asia	Iran	IRN	
7	62137802	Europe	Italy	ITA	
8	126451398	Asia	Japan	JPN	
9	16204486	Asia	Cambodia	KHM	
10	51181299	Asia	South Korea	KOR	
11	7126706	Asia	Laos	LAO	
12	22409381	Asia	Sri Lanka	LKA	
13	55123814	Asia	Myanmar	MMR	
14	31381992	Asia	Malaysia	MYS	
15	29384297	Asia	Nepal	NPL	
16	204924861	Asia	Pakistan	PAK	
17	104256076	Asia	Philippines	PHL	
18	25248140	Asia	North Korea	PRK	
19	68414135	Asia	Thailand	THA	

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20	23508428	Asia	Taiwan	TWN		
21	326625791	North America	United States of America	USA		
22	96160163	Asia	Vietnam	VNM		
gdp_md_est geometry \						
0	628400.0	POLYGON ((92.67272 22.04124, 92.65226 21.32405...				
1	3081000.0	POLYGON ((-53.37366 -33.76838, -53.65054 -33.2...				
2	21140000.0	MULTIPOLYGON (((109.47521 18.19770, 108.65521 ...				
3	1690000.0	POLYGON ((-7.45373 37.09779, -7.53711 37.42890...				
4	3028000.0	MULTIPOLYGON (((141.00021 -2.60015, 141.01706 ...				
5	8721000.0	POLYGON ((97.32711 28.26158, 97.40256 27.88254...				
6	1459000.0	POLYGON ((48.56797 29.92678, 48.01457 30.45246...				
7	2221000.0	MULTIPOLYGON (((10.44270 46.89355, 11.04856 46...				
8	4932000.0	MULTIPOLYGON (((141.88460 39.18086, 140.95949 ...				
9	58940.0	POLYGON ((102.58493 12.18659, 102.34810 13.394...				
10	1929000.0	POLYGON ((126.17476 37.74969, 126.23734 37.840...				
11	40960.0	POLYGON ((107.38273 14.20244, 106.49637 14.570...				
12	236700.0	POLYGON ((81.78796 7.52306, 81.63732 6.48178, ...				
13	311100.0	POLYGON ((100.11599 20.41785, 99.54331 20.1866...				
14	863000.0	MULTIPOLYGON (((100.08576 6.46449, 100.25960 6...				
15	71520.0	POLYGON ((88.12044 27.87654, 88.04313 27.44582...				
16	988200.0	POLYGON ((77.83745 35.49401, 76.87172 34.65354...				
17	801900.0	MULTIPOLYGON (((120.83390 12.70450, 120.32344 ...				
18	40000.0	MULTIPOLYGON (((130.78000 42.22001, 130.78000 ...				
19	1161000.0	POLYGON ((105.21878 14.27321, 104.28142 14.416...				
20	1127000.0	POLYGON ((121.77782 24.39427, 121.17563 22.790...				
21	18560000.0	MULTIPOLYGON (((-122.84000 49.00000, -120.0000...				
22	594900.0	POLYGON ((104.33433 10.48654, 105.19991 10.889...				
country_name tCH4_2015 tCH4_2016 \						
0	Bangladesh	2.344420e+06	2.278158e+06			
1	Brazil	3.410233e+05	3.104189e+05			
2	China	6.133647e+06	5.859531e+06			
3	Spain	1.141464e+04	1.334803e+04			
4	Indonesia	1.283649e+06	1.023129e+06			
5	India	6.219887e+06	5.309413e+06			
6	Iran (Islamic Republic of)	8.774407e+04	9.180121e+04			
7	Italy	4.995968e+04	4.937785e+04			
8	Japan	2.305465e+05	2.284133e+05			
9	Cambodia	4.954698e+05	5.731698e+05			
10	Korea (the Republic of)	1.451878e+05	1.274597e+05			
11	Lao People's Democratic Republic (the)	1.661169e+04	1.696441e+04			
12	Sri Lanka	8.305626e+04	1.011743e+05			
13	Myanmar	1.132082e+06	1.290806e+06			
14	Malaysia	1.057399e+05	1.110049e+05			
15	Nepal	1.007479e+05	6.667161e+04			
16	Pakistan	4.852431e+05	5.945922e+05			
17	Philippines (the)	3.432021e+05	4.073554e+05			
18	Korea (the Democratic People's Republic of)	1.143217e+05	9.177653e+04			
19	Thailand	1.393798e+06	1.780993e+06			
20	Taiwan (Province of China)	7.866956e+04	8.089149e+04			
21	United States of America (the)	1.611324e+05	1.618576e+05			
22	Viet Nam	1.346013e+06	1.483777e+06			
tCH4_2017 ... tCH4_2021 tCO2_2015 tCO2_2016 tCO2_2017 \						
0	2.098958e+06	...	1.983974e+06	5.861049e+07	5.695395e+07	5.247394e+07

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1	3.725173e+05	...	4.544874e+05	8.525583e+06	7.760473e+06	9.312934e+06
2	6.355071e+06	...	6.068210e+06	1.533412e+08	1.464883e+08	1.588768e+08
3	1.217299e+04	...	8.531579e+03	2.853661e+05	3.337007e+05	3.043248e+05
4	9.615327e+05	...	1.009936e+06	3.209122e+07	2.557824e+07	2.403832e+07
5	6.228451e+06	...	6.567960e+06	1.554972e+08	1.327353e+08	1.557113e+08
6	9.620217e+04	...	9.053525e+04	2.193602e+06	2.295030e+06	2.405054e+06
7	5.443679e+04	...	5.089759e+04	1.248992e+06	1.234446e+06	1.360920e+06
8	2.708935e+05	...	1.574007e+05	5.763662e+06	5.710333e+06	6.772337e+06
9	4.517045e+05	...	5.644891e+05	1.238675e+07	1.432925e+07	1.129261e+07
10	1.463222e+05	...	1.013006e+05	3.629695e+06	3.186493e+06	3.658056e+06
11	1.168063e+04	...	1.475014e+04	4.152924e+05	4.241102e+05	2.920158e+05
12	5.911841e+04	...	8.466966e+04	2.076407e+06	2.529358e+06	1.477960e+06
13	1.205169e+06	...	1.289837e+06	2.830206e+07	3.227014e+07	3.012923e+07
14	1.111291e+05	...	1.069696e+05	2.643498e+06	2.775123e+06	2.778227e+06
15	8.081300e+04	...	4.811408e+04	2.518697e+06	1.666790e+06	2.020325e+06
16	5.372641e+05	...	4.849205e+05	1.213108e+07	1.486480e+07	1.343160e+07
17	3.836830e+05	...	4.383270e+05	8.580052e+06	1.018389e+07	9.592074e+06
18	1.085457e+05	...	7.735988e+04	2.858041e+06	2.294413e+06	2.713641e+06
19	1.164699e+06	...	8.528673e+05	3.484495e+07	4.452483e+07	2.911748e+07
20	8.705634e+04	...	6.619861e+04	1.966739e+06	2.022287e+06	2.176408e+06
21	1.684799e+05	...	1.634842e+05	4.028310e+06	4.046440e+06	4.211999e+06
22	1.406437e+06	...	1.502787e+06	3.365033e+07	3.709441e+07	3.516092e+07

	tCO2_2018	tCO2_2019	Mean_CH4	Total_CH4	Mean_CO2 \
0	5.353076e+07	5.177463e+07	2.186750e+06	1.093375e+07	5.466875e+07
1	9.292575e+06	8.236783e+06	3.450268e+05	1.725134e+06	8.625670e+06
2	1.353491e+08	1.400838e+08	5.873113e+06	2.936556e+07	1.468278e+08
3	3.513524e+05	2.870810e+05	1.249460e+04	6.247300e+04	3.123650e+05
4	2.942454e+07	3.166670e+07	1.142392e+06	5.711960e+06	2.855980e+07
5	1.647450e+08	1.875389e+08	6.369821e+06	3.184910e+07	1.592455e+08
6	2.218936e+06	2.375050e+06	9.190138e+04	4.595069e+05	2.297534e+06
7	1.117475e+06	1.141729e+06	4.882850e+04	2.441425e+05	1.220712e+06
8	3.870631e+06	5.830141e+06	2.235768e+05	1.117884e+06	5.589421e+06
9	1.398153e+07	1.486819e+07	5.348666e+05	2.674333e+06	1.337166e+07
10	3.233858e+06	3.319455e+06	1.362205e+05	6.811023e+05	3.405512e+06
11	2.524186e+05	3.652645e+05	1.399281e+04	6.996406e+04	3.498203e+05
12	2.254728e+06	2.119022e+06	8.365981e+04	4.182990e+05	2.091495e+06
13	3.431117e+07	3.142221e+07	1.251478e+06	6.257392e+06	3.128696e+07
14	2.666313e+06	2.640717e+06	1.080310e+05	5.401551e+05	2.700775e+06
15	2.300188e+06	2.910588e+06	9.133271e+04	4.566635e+05	2.283318e+06
16	1.133074e+07	1.632137e+07	5.446368e+05	2.723184e+06	1.361592e+07
17	1.043803e+07	8.961374e+06	3.820433e+05	1.910216e+06	9.551082e+06
18	2.165645e+06	2.413765e+06	9.956405e+04	4.978202e+05	2.489101e+06
19	2.291644e+07	3.262615e+07	1.312239e+06	6.561194e+06	3.280597e+07
20	2.034538e+06	2.247717e+06	8.358152e+04	4.179076e+05	2.089538e+06
21	4.143136e+06	4.228377e+06	1.652661e+05	8.263305e+05	4.131652e+06
22	3.293637e+07	3.174377e+07	1.364686e+06	6.823432e+06	3.411716e+07

	Total_CO2
0	2.733438e+08
1	4.312835e+07
2	7.341391e+08
3	1.561825e+06
4	1.427990e+08
5	7.962276e+08
6	1.148767e+07

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```

7  6.103562e+06
8  2.794710e+07
9  6.685832e+07
10 1.702756e+07
11 1.749102e+06
12 1.045748e+07
13 1.564348e+08
14 1.350388e+07
15 1.141659e+07
16 6.807960e+07
17 4.775541e+07
18 1.244551e+07
19 1.640298e+08
20 1.044769e+07
21 2.065826e+07
22 1.705858e+08

```

```
[23 rows x 23 columns]
```

```
trace_merged_df = trace_merged_df[trace_merged_df['continent'] != 'Antarctica'].copy()
```

```

trace_merged_df
#trace_merged_df.dropna(axis=1, how='any')

```

	pop_est	continent	name	iso3_country	\
0	157826578	Asia	Bangladesh	BGD	
1	207353391	South America	Brazil	BRA	
2	1379302771	Asia	China	CHN	
3	48958159	Europe	Spain	ESP	
4	260580739	Asia	Indonesia	IDN	
5	1281935911	Asia	India	IND	
6	82021564	Asia	Iran	IRN	
7	62137802	Europe	Italy	ITA	
8	126451398	Asia	Japan	JPN	
9	16204486	Asia	Cambodia	KHM	
10	51181299	Asia	South Korea	KOR	
11	7126706	Asia	Laos	LAO	
12	22409381	Asia	Sri Lanka	LKA	
13	55123814	Asia	Myanmar	MMR	
14	31381992	Asia	Malaysia	MYS	
15	29384297	Asia	Nepal	NPL	
16	204924861	Asia	Pakistan	PAK	
17	104256076	Asia	Philippines	PHL	
18	25248140	Asia	North Korea	PRK	
19	68414135	Asia	Thailand	THA	
20	23508428	Asia	Taiwan	TWN	
21	326625791	North America	United States of America	USA	
22	96160163	Asia	Vietnam	VNM	

	gdp_md_est	geometry	\
0	628400.0	POLYGON ((92.67272 22.04124, 92.65226 21.32405...	
1	3081000.0	POLYGON ((-53.37366 -33.76838, -53.65054 -33.2...	
2	21140000.0	MULTIPOLYGON (((109.47521 18.19770, 108.65521 ...	
3	1690000.0	POLYGON ((-7.45373 37.09779, -7.53711 37.42890...	

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```

4 3028000.0 MULTIPOLYGON (((141.00021 -2.60015, 141.01706 ...
5 8721000.0 POLYGON ((97.32711 28.26158, 97.40256 27.88254...
6 1459000.0 POLYGON ((48.56797 29.92678, 48.01457 30.45246...
7 2221000.0 MULTIPOLYGON (((10.44270 46.89355, 11.04856 46...
8 4932000.0 MULTIPOLYGON (((141.88460 39.18086, 140.95949 ...
9 58940.0 POLYGON ((102.58493 12.18659, 102.34810 13.394...
10 1929000.0 POLYGON ((126.17476 37.74969, 126.23734 37.840...
11 40960.0 POLYGON ((107.38273 14.20244, 106.49637 14.570...
12 236700.0 POLYGON ((81.78796 7.52306, 81.63732 6.48178, ...
13 311100.0 POLYGON ((100.11599 20.41785, 99.54331 20.1866...
14 863000.0 MULTIPOLYGON (((100.08576 6.46449, 100.25960 6...
15 71520.0 POLYGON ((88.12044 27.87654, 88.04313 27.44582...
16 988200.0 POLYGON ((77.83745 35.49401, 76.87172 34.65354...
17 801900.0 MULTIPOLYGON (((120.83390 12.70450, 120.32344 ...
18 40000.0 MULTIPOLYGON (((130.78000 42.22001, 130.78000 ...
19 1161000.0 POLYGON ((105.21878 14.27321, 104.28142 14.416...
20 1127000.0 POLYGON ((121.77782 24.39427, 121.17563 22.790...
21 18560000.0 MULTIPOLYGON (((-122.84000 49.00000, -120.0000...
22 594900.0 POLYGON ((104.33433 10.48654, 105.19991 10.889...

```

	country_name	tCH4_2015	tCH4_2016 \
0	Bangladesh	2.344420e+06	2.278158e+06
1	Brazil	3.410233e+05	3.104189e+05
2	China	6.133647e+06	5.859531e+06
3	Spain	1.141464e+04	1.334803e+04
4	Indonesia	1.283649e+06	1.023129e+06
5	India	6.219887e+06	5.309413e+06
6	Iran (Islamic Republic of)	8.774407e+04	9.180121e+04
7	Italy	4.995968e+04	4.937785e+04
8	Japan	2.305465e+05	2.284133e+05
9	Cambodia	4.954698e+05	5.731698e+05
10	Korea (the Republic of)	1.451878e+05	1.274597e+05
11	Lao People's Democratic Republic (the)	1.661169e+04	1.696441e+04
12	Sri Lanka	8.305626e+04	1.011743e+05
13	Myanmar	1.132082e+06	1.290806e+06
14	Malaysia	1.057399e+05	1.110049e+05
15	Nepal	1.007479e+05	6.667161e+04
16	Pakistan	4.852431e+05	5.945922e+05
17	Philippines (the)	3.432021e+05	4.073554e+05
18	Korea (the Democratic People's Republic of)	1.143217e+05	9.177653e+04
19	Thailand	1.393798e+06	1.780993e+06
20	Taiwan (Province of China)	7.866956e+04	8.089149e+04
21	United States of America (the)	1.611324e+05	1.618576e+05
22	Viet Nam	1.346013e+06	1.483777e+06

	tCH4_2017 ...	tCH4_2021	tCO2_2015	tCO2_2016	tCO2_2017 \
0	2.098958e+06 ...	1.983974e+06	5.861049e+07	5.695395e+07	5.247394e+07
1	3.725173e+05 ...	4.544874e+05	8.525583e+06	7.760473e+06	9.312934e+06
2	6.355071e+06 ...	6.068210e+06	1.533412e+08	1.464883e+08	1.588768e+08
3	1.217299e+04 ...	8.531579e+03	2.853661e+05	3.337007e+05	3.043248e+05
4	9.615327e+05 ...	1.009936e+06	3.209122e+07	2.557824e+07	2.403832e+07
5	6.228451e+06 ...	6.567960e+06	1.554972e+08	1.327353e+08	1.557113e+08
6	9.620217e+04 ...	9.053525e+04	2.193602e+06	2.295030e+06	2.405054e+06
7	5.443679e+04 ...	5.089759e+04	1.248992e+06	1.234446e+06	1.360920e+06
8	2.708935e+05 ...	1.574007e+05	5.763662e+06	5.710333e+06	6.772337e+06
9	4.517045e+05 ...	5.644891e+05	1.238675e+07	1.432925e+07	1.129261e+07

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10	1.463222e+05	...	1.013006e+05	3.629695e+06	3.186493e+06	3.658056e+06
11	1.168063e+04	...	1.475014e+04	4.152924e+05	4.241102e+05	2.920158e+05
12	5.911841e+04	...	8.466966e+04	2.076407e+06	2.529358e+06	1.477960e+06
13	1.205169e+06	...	1.289837e+06	2.830206e+07	3.227014e+07	3.012923e+07
14	1.111291e+05	...	1.069696e+05	2.643498e+06	2.775123e+06	2.778227e+06
15	8.081300e+04	...	4.811408e+04	2.518697e+06	1.666790e+06	2.020325e+06
16	5.372641e+05	...	4.849205e+05	1.213108e+07	1.486480e+07	1.343160e+07
17	3.836830e+05	...	4.383270e+05	8.580052e+06	1.018389e+07	9.592074e+06
18	1.085457e+05	...	7.735988e+04	2.858041e+06	2.294413e+06	2.713641e+06
19	1.164699e+06	...	8.528673e+05	3.484495e+07	4.452483e+07	2.911748e+07
20	8.705634e+04	...	6.619861e+04	1.966739e+06	2.022287e+06	2.176408e+06
21	1.684799e+05	...	1.634842e+05	4.028310e+06	4.046440e+06	4.211999e+06
22	1.406437e+06	...	1.502787e+06	3.365033e+07	3.709441e+07	3.516092e+07

	tCO2_2018	tCO2_2019	Mean_CH4	Total_CH4	Mean_CO2 \
0	5.353076e+07	5.177463e+07	2.186750e+06	1.093375e+07	5.466875e+07
1	9.292575e+06	8.236783e+06	3.450268e+05	1.725134e+06	8.625670e+06
2	1.353491e+08	1.400838e+08	5.873113e+06	2.936556e+07	1.468278e+08
3	3.513524e+05	2.870810e+05	1.249460e+04	6.247300e+04	3.123650e+05
4	2.942454e+07	3.166670e+07	1.142392e+06	5.711960e+06	2.855980e+07
5	1.647450e+08	1.875389e+08	6.369821e+06	3.184910e+07	1.592455e+08
6	2.218936e+06	2.375050e+06	9.190138e+04	4.595069e+05	2.297534e+06
7	1.117475e+06	1.141729e+06	4.882850e+04	2.441425e+05	1.220712e+06
8	3.870631e+06	5.830141e+06	2.235768e+05	1.117884e+06	5.589421e+06
9	1.398153e+07	1.486819e+07	5.348666e+05	2.674333e+06	1.337166e+07
10	3.233858e+06	3.319455e+06	1.362205e+05	6.811023e+05	3.405512e+06
11	2.524186e+05	3.652645e+05	1.399281e+04	6.996406e+04	3.498203e+05
12	2.254728e+06	2.119022e+06	8.365981e+04	4.182990e+05	2.091495e+06
13	3.431117e+07	3.142221e+07	1.251478e+06	6.257392e+06	3.128696e+07
14	2.666313e+06	2.640717e+06	1.080310e+05	5.401551e+05	2.700775e+06
15	2.300188e+06	2.910588e+06	9.133271e+04	4.566635e+05	2.283318e+06
16	1.133074e+07	1.632137e+07	5.446368e+05	2.723184e+06	1.361592e+07
17	1.043803e+07	8.961374e+06	3.820433e+05	1.910216e+06	9.551082e+06
18	2.165645e+06	2.413765e+06	9.956405e+04	4.978202e+05	2.489101e+06
19	2.291644e+07	3.262615e+07	1.312239e+06	6.561194e+06	3.280597e+07
20	2.034538e+06	2.247717e+06	8.358152e+04	4.179076e+05	2.089538e+06
21	4.143136e+06	4.228377e+06	1.652661e+05	8.263305e+05	4.131652e+06
22	3.293637e+07	3.174377e+07	1.364686e+06	6.823432e+06	3.411716e+07

	Total_CO2
0	2.733438e+08
1	4.312835e+07
2	7.341391e+08
3	1.561825e+06
4	1.427990e+08
5	7.962276e+08
6	1.148767e+07
7	6.103562e+06
8	2.794710e+07
9	6.685832e+07
10	1.702756e+07
11	1.749102e+06
12	1.045748e+07
13	1.564348e+08
14	1.350388e+07
15	1.141659e+07

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```

16  6.807960e+07
17  4.775541e+07
18  1.244551e+07
19  1.640298e+08
20  1.044769e+07
21  2.065826e+07
22  1.705858e+08

[23 rows x 23 columns]

```

### 6.3.2 Trace Merge to File

```

gisout = "/Users/jnapolitano/Projects/wattime-takehome/data/TRACE_DATA_GEO.geojson"
trace_merged_df.to_file(gisout, driver="GeoJSON")

```

```

/Users/jnapolitano/venvs/finance/lib/python3.9/site-packages/geopandas/io/file.
py:362: FutureWarning: pandas.Int64Index is deprecated and will be removed from
pandas in a future version. Use pandas.Index with the appropriate dtype instead.
pd.Int64Index,

```

## 6.4 Merge GEO DATA

```

filepath = "/Users/jnapolitano/Projects/wattime-takehome/data/MERGED_DATA.csv"

merged_data = pd.read_csv(filepath)

merged_data.drop('Unnamed: 0', axis = 1 , inplace = True)

```

```

merged_countries = world.merge(merged_data, on='iso3_country', how='right', sort=True)

```

```
merged_countries
```

	pop_est	continent	name	iso3_country \
0	1.578266e+08	Asia	Bangladesh	BGD
1	2.073534e+08	South America	Brazil	BRA
2	1.379303e+09	Asia	China	CHN
3	4.895816e+07	Europe	Spain	ESP
4	2.605807e+08	Asia	Indonesia	IDN
5	1.281936e+09	Asia	India	IND
6	8.202156e+07	Asia	Iran	IRN
7	6.213780e+07	Europe	Italy	ITA
8	1.264514e+08	Asia	Japan	JPN
9	1.620449e+07	Asia	Cambodia	KHM
10	5.118130e+07	Asia	South Korea	KOR
11	7.126706e+06	Asia	Laos	LAO
12	2.240938e+07	Asia	Sri Lanka	LKA
13	5.512381e+07	Asia	Myanmar	MMR

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14	3.138199e+07	Asia	Malaysia	MYS
15	2.938430e+07	Asia	Nepal	NPL
16	2.049249e+08	Asia	Pakistan	PAK
17	1.042561e+08	Asia	Philippines	PHL
18	2.524814e+07	Asia	North Korea	PRK
19	6.841414e+07	Asia	Thailand	THA
20	2.350843e+07	Asia	Taiwan	TWN
21	3.266258e+08	North America	United States of America	USA
22	9.616016e+07	Asia	Vietnam	VNM
23	NaN	NaN	NaN	NaN
24	NaN	NaN	NaN	NaN

	gdp_md_est	geometry	\
0	628400.0	POLYGON ((92.67272 22.04124, 92.65226 21.32405...	
1	3081000.0	POLYGON ((-53.37366 -33.76838, -53.65054 -33.2...	
2	21140000.0	MULTIPOLYGON (((109.47521 18.19770, 108.65521 ...	
3	1690000.0	POLYGON ((-7.45373 37.09779, -7.53711 37.42890...	
4	3028000.0	MULTIPOLYGON (((141.00021 -2.60015, 141.01706 ...	
5	8721000.0	POLYGON ((97.32711 28.26158, 97.40256 27.88254...	
6	1459000.0	POLYGON ((48.56797 29.92678, 48.01457 30.45246...	
7	2221000.0	MULTIPOLYGON (((10.44270 46.89355, 11.04856 46...	
8	4932000.0	MULTIPOLYGON (((141.88460 39.18086, 140.95949 ...	
9	58940.0	POLYGON ((102.58493 12.18659, 102.34810 13.394...	
10	1929000.0	POLYGON ((126.17476 37.74969, 126.23734 37.840...	
11	40960.0	POLYGON ((107.38273 14.20244, 106.49637 14.570...	
12	236700.0	POLYGON ((81.78796 7.52306, 81.63732 6.48178, ...	
13	311100.0	POLYGON ((100.11599 20.41785, 99.54331 20.1866...	
14	863000.0	MULTIPOLYGON (((100.08576 6.46449, 100.25960 6...	
15	71520.0	POLYGON ((88.12044 27.87654, 88.04313 27.44582...	
16	988200.0	POLYGON ((77.83745 35.49401, 76.87172 34.65354...	
17	801900.0	MULTIPOLYGON (((120.83390 12.70450, 120.32344 ...	
18	40000.0	MULTIPOLYGON (((130.78000 42.22001, 130.78000 ...	
19	1161000.0	POLYGON ((105.21878 14.27321, 104.28142 14.416...	
20	1127000.0	POLYGON ((121.77782 24.39427, 121.17563 22.790...	
21	18560000.0	MULTIPOLYGON (((-122.84000 49.00000, -120.0000...	
22	594900.0	POLYGON ((104.33433 10.48654, 105.19991 10.889...	
23	NaN		None
24	NaN		None

	country_name_FAOSTAT	2015	2016	\
0	Bangladesh	1131293.4	1.093480e+06	
1	Brazil	138910.3	1.262782e+05	
2	China	5406593.9	5.399920e+06	
3	Spain	55082.2	5.507310e+04	
4	Indonesia	2407953.5	2.387656e+06	
5	India	4580248.4	4.559136e+06	
6	Iran (Islamic Republic of)	116486.7	1.310085e+05	
7	Italy	114574.8	1.180030e+05	
8	Japan	330353.1	3.264030e+05	
9	Cambodia	436826.0	4.590031e+05	
10	Korea (the Republic of)	167862.2	1.635341e+05	
11	Lao People's Democratic Republic (the)	94826.8	9.563000e+04	
12	Sri Lanka	132640.0	1.217563e+05	
13	Myanmar	1059409.6	1.052288e+06	
14	Malaysia	121942.6	1.232328e+05	
15	Nepal	149262.2	1.427237e+05	

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16		Pakistan	383529.3	3.813618e+05
17		Philippines (the)	1557810.6	1.524292e+06
18		Korea (the Democratic People's Republic of)	82823.3	8.344230e+04
19		Thailand	1554254.0	1.703328e+06
20		Taiwan (Province of China)	45838.7	4.983830e+04
21		United States of America (the)	364728.0	4.386620e+05
22		Viet Nam	1381744.4	1.365174e+06
23		mean	948478.0	9.522272e+05
24		total	21814994.0	2.190123e+07

	2017	...	CO2_abs_percent_diff_2019	CO2_abs_percent_diff_means	\
0	1.154531e+06	...	57.606797	63.425917	
1	1.303229e+05	...	99.141295	93.222402	
2	5.400129e+06	...	7.189945	9.421813	
3	5.423240e+04	...	127.757634	124.705066	
4	2.425291e+06	...	56.234955	70.153790	
5	4.620791e+06	...	47.515394	32.086209	
6	8.723360e+04	...	1.152675	13.260902	
7	1.180030e+05	...	83.321649	80.187551	
8	3.237003e+05	...	31.555853	36.874729	
9	4.737453e+05	...	23.769728	14.304565	
10	1.584897e+05	...	14.321600	15.813819	
11	9.394070e+04	...	136.209544	145.627378	
12	8.445630e+04	...	18.612972	27.569003	
13	1.087030e+06	...	14.853751	14.697380	
14	1.226563e+05	...	14.753544	13.043224	
15	1.625746e+05	...	29.190154	50.437328	
16	4.060833e+05	...	42.334367	31.154530	
17	1.609862e+06	...	125.114417	121.748165	
18	8.459620e+04	...	15.169422	17.492669	
19	1.714466e+06	...	17.404685	22.551331	
20	4.999120e+04	...	58.617135	52.458032	
21	3.362555e+05	...	69.713577	78.850049	
22	1.360552e+06	...	3.761740	0.902428	
23	9.590840e+05	...	47.621862	49.129925	
24	2.205893e+07	...	1095.302835	1129.988279	

	CO2_abs_percent_diff_totals	CO2_relative_percent_diff_2015	\
0	63.425917	-107.233570	
1	93.222402	-145.498958	
2	9.421813	-13.447527	
3	124.705066	79.277072	
4	70.153790	46.691296	
5	32.086209	-35.798018	
6	13.260902	24.674600	
7	80.187551	56.395576	
8	36.874729	30.212099	
9	14.304565	-13.424988	
10	15.813819	13.507737	
11	145.627378	82.482067	
12	27.569003	37.382189	
13	14.697380	-6.859731	
14	13.043224	13.287137	
15	50.437328	32.502761	
16	31.154530	-26.520489	
17	121.748165	77.968948	

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18	17.492669	-38.030793
19	22.551331	10.323662
20	52.458032	-71.622589
21	78.850049	55.821215
22	0.902428	2.585962
23	49.129925	4.551116
24	1129.988279	104.675658
CO2_relative_percent_diff_2016 CO2_relative_percent_diff_2017 \		
0	-108.340089	-81.801753
1	-145.821472	-185.841820
2	-8.511442	-17.683689
3	75.763074	77.554025
4	57.149216	60.353916
5	-16.456550	-34.791882
6	29.927285	-10.281098
7	58.155426	53.868303
8	30.021068	16.313490
9	-24.872761	4.652461
10	22.059230	7.677136
11	82.260369	87.565950
12	16.904226	30.001182
13	-22.666619	-10.868129
14	9.922595	9.398001
15	53.286238	50.291743
16	-55.912882	-32.303912
17	73.275770	76.166725
18	-9.988016	-28.310326
19	-4.559627	32.066348
20	-62.307893	-74.143329
21	63.101978	49.895259
22	-8.687738	-3.372544
23	4.508756	3.322003
24	103.701389	76.406059
CO2_relative_percent_diff_2018 CO2_relative_percent_diff_2019 \		
0	-87.073857	-80.912284
1	-205.638590	-196.594424
2	-2.108366	-7.458060
3	73.445258	77.958601
4	51.073546	43.893273
5	-41.376939	-62.321563
6	5.513463	1.146070
7	59.165480	58.817707
8	51.954183	27.255501
9	-16.667610	-26.975760
10	16.497815	13.364589
11	87.883944	81.026578
12	18.784377	17.028241
13	-22.665852	-16.045425
14	14.840462	13.739977
15	40.212459	25.472433
16	-15.207139	-53.701452
17	74.003201	76.966391
18	-3.195586	-16.414407
19	46.173612	16.011325

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```

20          -64.692890          -82.919716
21          59.792699          51.694526
22          1.405162           3.692293
23          6.179080          -1.533721
24         142.118831          -35.275585

      CO2_relative_percent_diff_means  CO2_relative_percent_diff_totals
0                -92.881337                -92.881337
1             -174.610414             -174.610414
2              -9.887609              -9.887609
3              76.811284              76.811284
4              51.936188              51.936188
5             -38.217479             -38.217479
6              12.436318              12.436318
7              57.238482              57.238482
8              31.134372              31.134372
9             -15.406481             -15.406481
10             14.655057             14.655057
11             84.268427             84.268427
12             24.229137             24.229137
13            -15.863111            -15.863111
14             12.244674             12.244674
15             40.279401             40.279401
16            -36.903009            -36.903009
17             75.679166             75.679166
18            -19.169278            -19.169278
19             20.266184             20.266184
20            -71.109302            -71.109302
21             56.553728             56.553728
22             -0.906518             -0.906518
23             3.599038             3.599038
24             82.777881             82.777881

[25 rows x 78 columns]

```

### 6.4.1 Merged Countries to File

```
gisout = "/Users/jnapolitano/Projects/watttime-takehome/data/MERGE_DATA_GEO.geojson"
merged_countries.to_file(gisout, driver="GeoJSON")
```

```

/Users/jnapolitano/venvs/finance/lib/python3.9/site-packages/geopandas/io/file.
py:362: FutureWarning: pandas.Int64Index is deprecated and will be removed from
pandas in a future version. Use pandas.Index with the appropriate dtype instead.
pd.Int64Index,

```