## fish-mercury-data-analysis

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### 1 Fish mercury data analysis

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Mercury contamination of fish in edible freshwater is a direct threat to our health. A recent study was conducted in 53 Florida lakes to examine factors influencing the level of mercury contamination. The variables that were measured can be found in mercury.csv and are described as follows:

X1 = identification number X2 = lake name X3 = alkalinity (mg/l calcium carbonate) X4 = PH X5 = calcium (mg/l) X6 = chlorophyll (mg/l) X7 = average mercury concentration (part per million) in the muscle tissue of the group of fish studied in each lake X8 = number of fish studied in each lake X8 = number of fish studied in the lake X9 = minimum mercury concentration in each group of fish X10 = maximum mercury concentration in each group of fish X10 = maximum mercury concentration in each group of fish X11 = estimate (by regression) of mercury concentration in the 3-year-old fish (or average mercury when age is not available) X12 = indicator of the age of the fish (0: young; 1: mature).

Around the main research question that arises in this study: what are the main factors influencing the level of mercury contamination in fish in Florida lakes? Parallel questions may arise that break down this general question:

- Is there evidence to assume that the average concentration of mercury in the lakes is harmful to human health? Consider that the reference regulations for assessing maximum Hg levels (Regulation 34687-MAG and international regulations EC 1881/2006 and Codex Standard 193-1995) state that the average mercury concentration in fishery products should not exceed 0.5 mg Hg/kg.
- Will there be significant difference between mercury concentration by age of fish?
- If the sampling was done by casting a net and analyzing the fish that the net encountered, will the number of fish encountered influence the mercury concentration in the fish?
- Do the concentrations of alkalinity, chlorophyll, calcium in the lake water influence the mercury concentration of the fish?

### 1.1 Database exploration

### 1.1.1 Accessing the database and setting up the required libraries

#### 1.1.2 Analyze dataset features

In the next codeblock we can see all the available features of the dataset already described

```
['name' 'alkalinity' 'ph' 'calcium' 'chlorophyll' 'mean_merc_porc'
  'num_fish' 'min_merc_porc' 'max_merc_porc' 'merc_estimate_porc'
  'age_fishes']
```

Which features are categorical? These values classify the samples into sets of similar samples. Within categorical features are the values nominal, ordinal, ratio, or interval based? Among other things this helps us select the appropriate plots for visualization.

• Nominal: name.

Which features are numerical? Which features are numerical? These values change from sample to sample. Within numerical features are the values discrete, continuous, or timeseries based? Among other things this helps us select the appropriate plots for visualization.

- Continuous: alkalinity, ph, calcium, chlorophyll, mean\_merc\_porc, min\_merc\_porc, max\_merc\_porc and merc\_estimate\_porc.
- Discrete: age\_fishes and num\_fish.

|    | name         | alkalinity  | ph    | calcium  | chlorophyll  | mean_m | erc_porc | \  |
|----|--------------|-------------|-------|----------|--------------|--------|----------|----|
| X1 |              |             |       |          |              |        |          |    |
| 1  | Alligator    | 5.9         | 6.1   | 3.0      | 0.7          |        | 1.23     |    |
| 2  | Annie        | 3.5         | 5.1   | 1.9      | 3.2          |        | 1.33     |    |
| 3  | Apopka       | 116.0       | 9.1   | 44.1     | 128.3        |        | 0.04     |    |
| 4  | Blue Cypress | 39.4        | 6.9   | 16.4     | 3.5          |        | 0.44     |    |
| 5  | Brick        | 2.5         | 4.6   | 2.9      | 1.8          |        | 1.20     |    |
|    |              |             |       |          |              |        |          |    |
|    | num_fish mi  | n_merc_porc | max_m | erc_porc | merc_estimat | e_porc | age_fish | es |
| X1 |              |             |       |          |              |        |          |    |
| 1  | 5            | 0.85        |       | 1.43     |              | 1.53   |          | 1  |
| 2  | 7            | 0.92        |       | 1.90     |              | 1.33   |          | 0  |
| 3  | 6            | 0.04        |       | 0.06     |              | 0.04   |          | 0  |
| 4  | 12           | 0.13        |       | 0.84     |              | 0.44   |          | 0  |
| 5  | 12           | 0.69        |       | 1.50     |              | 1.33   |          | 1  |

**Extra observations** The dataset have no missing values in any feature, which means we don't have to worry about fill empty values. We can observe that there it is 11 features that are numerical data type and 1 that are string data type.

We might need to convert in future the string features (categorical data) to numerical data types (numerical data).

<class 'pandas.core.frame.DataFrame'>
Int64Index: 53 entries, 1 to 53
Data columns (total 11 columns):

| # | Column     | Non-Null Count | Dtype   |
|---|------------|----------------|---------|
|   |            |                |         |
| 0 | name       | 53 non-null    | object  |
| 1 | alkalinity | 53 non-null    | float64 |
| 2 | ph         | 53 non-null    | float64 |

| 3  | calcium            | 53 non-null | float64 |
|----|--------------------|-------------|---------|
| 4  | chlorophyll        | 53 non-null | float64 |
| 5  | mean_merc_porc     | 53 non-null | float64 |
| 6  | num_fish           | 53 non-null | int64   |
| 7  | min_merc_porc      | 53 non-null | float64 |
| 8  | max_merc_porc      | 53 non-null | float64 |
| 9  | merc_estimate_porc | 53 non-null | float64 |
| 10 | age_fishes         | 53 non-null | int64   |
| _  |                    |             |         |

dtypes: float64(8), int64(2), object(1)

memory usage: 5.0+ KB

### 1.1.3 Database exploration

### Describe the numerical features

|       | alkalinity | ph        | calcium   | chlorophyll | mean_merc_porc | \ |
|-------|------------|-----------|-----------|-------------|----------------|---|
| count | 53.000000  | 53.000000 | 53.000000 | 53.000000   | 53.000000      |   |
| mean  | 37.530189  | 6.590566  | 22.201887 | 23.116981   | 0.527170       |   |
| std   | 38.203527  | 1.288449  | 24.932574 | 30.816321   | 0.341036       |   |
| min   | 1.200000   | 3.600000  | 1.100000  | 0.700000    | 0.040000       |   |
| 25%   | 6.600000   | 5.800000  | 3.300000  | 4.600000    | 0.270000       |   |
| 50%   | 19.600000  | 6.800000  | 12.600000 | 12.800000   | 0.480000       |   |
| 75%   | 66.500000  | 7.400000  | 35.600000 | 24.700000   | 0.770000       |   |
| max   | 128.000000 | 9.100000  | 90.700000 | 152.400000  | 1.330000       |   |
|       |            |           |           |             |                |   |

|       | ${\tt num\_fish}$ | min_merc_porc | max_merc_porc | merc_estimate_porc | age_fishes |
|-------|-------------------|---------------|---------------|--------------------|------------|
| count | 53.000000         | 53.000000     | 53.000000     | 53.000000          | 53.000000  |
| mean  | 13.056604         | 0.279811      | 0.874528      | 0.513208           | 0.811321   |
| std   | 8.560677          | 0.226406      | 0.522047      | 0.338729           | 0.394998   |
| min   | 4.000000          | 0.040000      | 0.060000      | 0.040000           | 0.000000   |
| 25%   | 10.000000         | 0.090000      | 0.480000      | 0.250000           | 1.000000   |
| 50%   | 12.000000         | 0.250000      | 0.840000      | 0.450000           | 1.000000   |
| 75%   | 12.000000         | 0.330000      | 1.330000      | 0.700000           | 1.000000   |
| max   | 44.000000         | 0.920000      | 2.040000      | 1.530000           | 1.000000   |

### Describe the categorical features

|        | name      |
|--------|-----------|
| count  | 53        |
| unique | 53        |
| top    | Alligator |
| freq   | 1         |

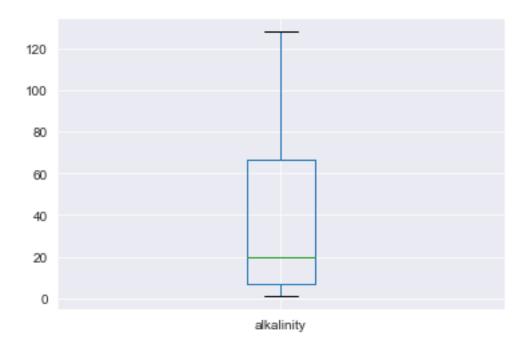
| name           | Alligator |
|----------------|-----------|
| alkalinity     | 17.3      |
| ph             | 5.8       |
| calcium        | 3.0       |
| chlorophyll    | 1.6       |
| mean_merc_porc | 0.34      |

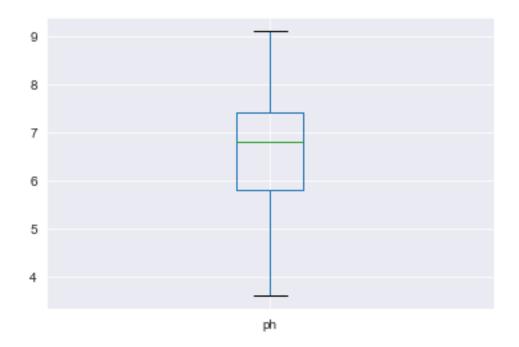
| num_fish           | 12.0 |
|--------------------|------|
| min_merc_porc      | 0.04 |
| max_merc_porc      | 0.06 |
| merc_estimate_porc | 0.16 |
| age_fishes         | 1.0  |
| N                  |      |

Name: 0, dtype: object

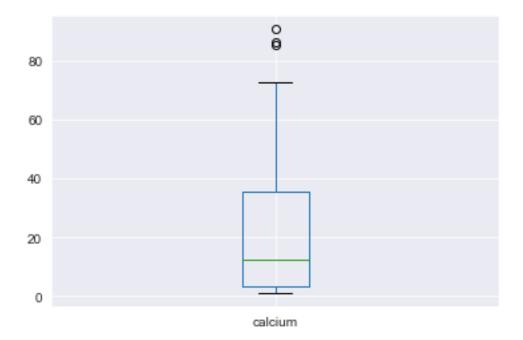
### Describe the variables with using visualization tools

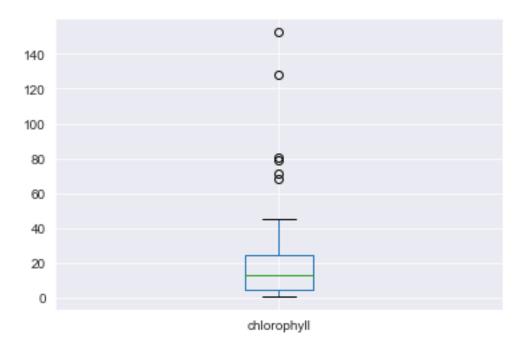
# Cuantitative variables positional measures and outliers detection <AxesSubplot:>



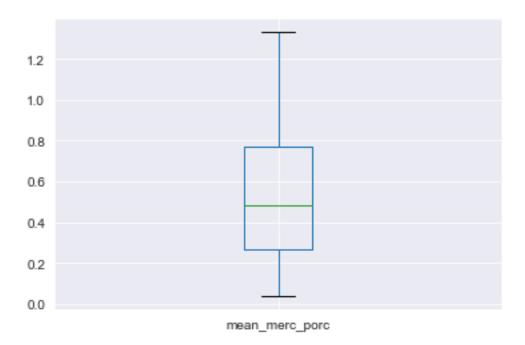


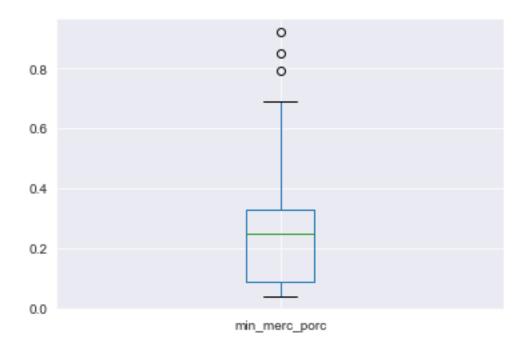
# <AxesSubplot:>



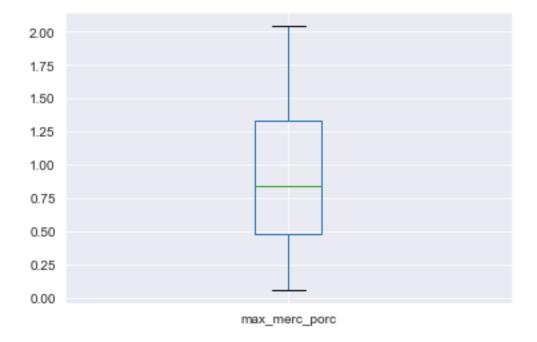


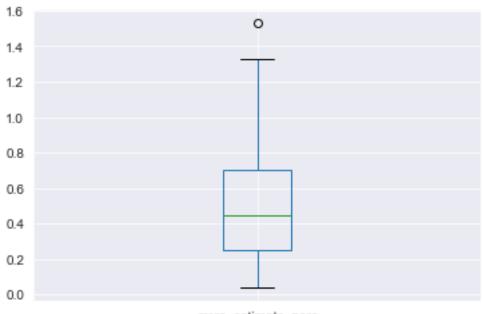
## <AxesSubplot:>



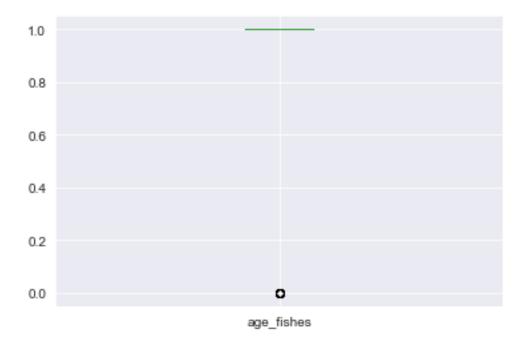


## <AxesSubplot:>

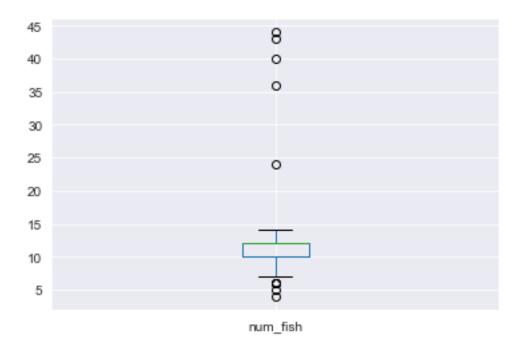




# merc\_estimate\_porc



<AxesSubplot:>



We get the salaries outliers:

```
X1
3 Apopka
9 Deer Point
14 East Tohopekaliga
17 Griffin
26 Kissimmee
Name: name, dtype: object
```

7

**Identify the data quality** Here we are looking for duplicated data, missing data and validating the relevance of each variable. There is no duplicated regiestries

0

There's is no null data

```
name 0
alkalinity 0
ph 0
calcium 0
chlorophyll 0
mean_merc_porc 0
num_fish 0
```

min\_merc\_porc 0
max\_merc\_porc 0
merc\_estimate\_porc 0
age\_fishes 0

dtype: int64