Hypothesis testing - Air Quality Index (AQI)

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1 Hypothesis testing - Air Quality Index (AQI)

1.1 Introduction

An environmental think tank called Repair Our Air (ROA). ROA is formulating policy recommendations to improve the air quality in America, using the Environmental Protection Agency's Air Quality Index (AQI) to guide their decision making. An AQI value close to 0 signals "little to no" public health concern, while higher values are associated with increased risk to public health.

They've tasked you with leveraging AQI data to help them prioritize their strategy for improving air quality in America.

ROA is considering the following decisions. For each, construct a hypothesis test and an accompanying visualization, using your results of that test to make a recommendation:

- 1. ROA is considering a metropolitan-focused approach. Within California, they want to know if the mean AQI in Los Angeles County is statistically different from the rest of California.
- 2. With limited resources, ROA has to choose between New York and Ohio for their next regional office. Does New York have a lower AQI than Ohio?
- 3. A new policy will affect those states with a mean AQI of 10 or greater. Will Michigan be affected by this new policy?

Notes: 1. 5% level of significance.

1.2 Step1 : Import Packages

```
[1]: import pandas as pd from scipy import stats
```

```
Load Dataset
```

```
[2]: df = pd.read_csv('c4_epa_air_quality.csv')
```

1.3 Step2: Data Exploration

```
[3]: df
[3]:
          Unnamed: 0
                       date_local
                                              state_name
                                                                     county_name
                       2018-01-01
                                                  Arizona
                                                                        Maricopa
     0
     1
                       2018-01-01
                                                     Ohio
                                                                         Belmont
     2
                       2018-01-01
                                                 Wyoming
                                                                           Teton
     3
                    3
                       2018-01-01
                                            Pennsylvania
                                                                    Philadelphia
     4
                       2018-01-01
                                                                            Polk
                                                     Iowa
     255
                  255
                       2018-01-01
                                   District Of Columbia
                                                           District of Columbia
     256
                  256
                                               Wisconsin
                       2018-01-01
                                                                           Dodge
     257
                  257
                       2018-01-01
                                                 Kentucky
                                                                       Jefferson
                  258
     258
                       2018-01-01
                                                 Nebraska
                                                                         Douglas
     259
                  259
                       2018-01-01
                                          North Carolina
                                                                            Wake
                                                              local_site_name \
              city_name
     0
                Buckeye
                                                                       BUCKEYE
     1
              Shadyside
                                                                     Shadyside
     2
          Not in a city
                          Yellowstone National Park - Old Faithful Snow ...
     3
           Philadelphia
                                                       North East Waste (NEW)
                                                                     CARPENTER
     4
             Des Moines
     . .
     255
             Washington
                                                                     Near Road
     256
                                                        HORICON WILDLIFE AREA
               Kekoskee
     257
             Louisville
                                                                  CANNONS LANE
     258
                  Omaha
                                                                           NaN
     259
                                                                    Triple Oak
          Not in a city
           parameter_name
                             units_of_measure
                                                arithmetic_mean
     0
          Carbon monoxide
                            Parts per million
                                                        0.473684
                                                                     7
     1
          Carbon monoxide
                            Parts per million
                                                        0.263158
                                                                     5
     2
                            Parts per million
                                                        0.111111
                                                                     2
          Carbon monoxide
     3
                            Parts per million
          Carbon monoxide
                                                        0.300000
                                                                     3
     4
                                                                     3
          Carbon monoxide
                            Parts per million
                                                        0.215789
                                                         •••
                                                                     3
     255
         Carbon monoxide
                            Parts per million
                                                        0.244444
          Carbon monoxide
                            Parts per million
                                                                     2
     256
                                                        0.200000
     257
          Carbon monoxide
                            Parts per million
                                                        0.163158
                                                                     2
     258
                                                                     9
          Carbon monoxide
                            Parts per million
                                                        0.421053
     259
          Carbon monoxide
                            Parts per million
                                                                     2
                                                        0.188889
     [260 rows x 10 columns]
    df.describe(include = 'all')
```

[4]:		Unnamed: 0	date_local		state_name		cour	county_name		city_	name	\	
	count	260.000000	260		260		260		260				
	unique	NaN		1		52		149			190		
	top	NaN	2018	3-01-01	Californ	nia	Los	Angeles	Not	in a	city		
	freq	NaN		260		66		14			21		
	mean	129.500000	NaN		N	IaN		NaN			NaN		
	std	75.199734		NaN	N	JaN		NaN			NaN		
	min	0.000000	NaN		NaN		NaN			NaN			
	25%	64.750000	NaN		N	IaN	NaN			NaN			
	50%	129.500000	NaN		NaN		NaN			NaN			
	75%	194.250000	NaN		NaN		NaN			NaN			
	max	259.000000	NaN		NaN		NaN			NaN			
		local_site_n	-		eter_name	u	\mathtt{nits}	ts_of_measure		arithm	etic_	mean	\
	count		257		260		260			260.000000			
	unique		253		1		1			NaN			
	top	Kapo	lei Carbor		${\tt monoxide}$	Pa	rts per million		.on	NaN		NaN	
	freq		2		260		260			NaN			
	mean		NaN		NaN		NaN		1aN	0.403169			
	std		NaN		NaN		NaN		1aN		0.31	7902	
	min		NaN		NaN		NaN			0.000000			
	25%		NaN		NaN			N	1aN		0.20	0000	
	50%		NaN		NaN			N	1aN		0.27	6315	
	75%		NaN		NaN			I.	1aN		0.51	6009	
	max		NaN		NaN			Ŋ	laN		1.92	1053	
		aqi											
	count	260.000000											
	unique	NaN											
	top	NaN											
	freq	NaN											
	mean	6.757692											
	std	7.061707											
	min	0.000000											
	25%	2.000000											
	50%	5.000000											
	75%	9.000000											
	max	50.000000											

[5]: df.shape

[5]: (260, 10)

Points from the preceding data exploration

- 1. California state has highest count among states
- 2. Los Angles city has highest count among counties

- 3. There are 52 states and 159 cities in the dataset
- 4. All the readings are on the same day.
- 5. Majority of reading are not in the cities.
- 6. Mean aqi is approx 6.5
- 7. 75 % of the reading are equal or less than 9 agi.

1.4 Step 3. Statistical Tests

- 1. Formulate the null hypothesis and the alternative hypothesis.
- 2. Set the significance level.
- 3. Determine the appropriate test procedure.
- 4. Compute the p-value.
- 5. Draw conclusion.
- 1.4.1 Hypothesis 1: ROA is considering a metropolitan-focused approach. Within California, they want to know if the mean AQI in Los Angeles County is statistically different from the rest of California.

```
[10]: # Create dataframes for each sample being compared

df_losangeles = df[df['county_name']=='Los Angeles']

df_california = df[(df['state_name']=='California') & (df['county_name']!='Los_

→Angeles')]
```

Formulate hypothesis: Formulate null and alternative hypotheses:

- H_0 : There is no difference in the mean AQI between Los Angeles County and the rest of California.
- H_A : There is a difference in the mean AQI between Los Angeles County and the rest of California.

Set the significance level:

```
[7]: # For this analysis, the significance level is 5% significance_level = 0.05
```

Determine the appropriate test procedure: For comparing the sample means between two independent samples, utilize a **two-sample -test**.

Compute the P-value

```
[12]: t_stat,p_val = stats.

→ttest_ind(df_losangeles['aqi'],df_california['aqi'],equal_var=False)
```

P-value for hypothesis 1: 0.049839056842410995
T-Statistic for hypothesis 1: 2.1107010796372014
Reject Null Hypothesis. There is a statistical evidence that there is difference in the mean AQI between Los Angeles County and the rest of California.

[]:

1.4.2 Hypothesis 2: With limited resources, ROA has to choose between New York and Ohio for their next regional office. Does New York have a lower AQI than Ohio?

```
[14]: # Create dataframes for each sample being compared
df_newyork = df[df['state_name'] == 'New York']
df_ohio = df[df['state_name'] == 'Ohio']
```

Formulate hypothesis: Formulate null and alternative hypotheses:

- H_0 : The mean AQI of New York is greater than or equal to that of Ohio.
- H_A : The mean AQI of New York is **below** that of Ohio.

Significance Level (remains at 5%)

Determine the appropriate test procedure: For comparing the sample means between two independent samples, utilize a **two-sample -test**.

Compute the P-value

```
[17]: print('P-value for hypothesis 2: ',p_val)
print('T-Statistic for hypothesis 2: ',t_stat)
```

```
if p_val <= significance_level:
    print('Reject Null Hypothesis. There is a statistical evidence that the
    →mean AQI of New York is below that of Ohio.')
else:
    print('Fail to reject Null Hypothesis. There not enough statistical
    →evidence that the mean AQI of New York is below that of Ohio')</pre>
```

P-value for hypothesis 2: 0.030446502691934697
T-Statistic for hypothesis 2: -2.025951038880333
Reject Null Hypothesis. There is a statistical evidence that the mean AQI of New York is below that of Ohio.

1.4.3 Hypothesis 3: A new policy will affect those states with a mean AQI of 10 or greater. Will Michigan be affected by this new policy?

```
[23]: # Create dataframes for each sample being compared df_michigan = df[df['state_name'] == 'Michigan']
```

Formulate your hypothesis: Formulate your null and alternative hypotheses here:

- H_0 : The mean AQI of Michigan is less than or equal to 10.
- H_A : The mean AQI of Michigan is greater than 10.

Significance Level (remains at 5%)

Determine the appropriate test procedure: comparing one sample mean relative to a particular value in one direction, utilize a one-sample -test.

Compute the P-value

```
[28]: t_stat, p_value = stats.ttest_1samp(df_michigan['aqi'], 10, u →alternative='greater')
```

```
[29]: print('P-value for hypothesis 3: ',p_val)
print('T-Statistic for hypothesis 3: ',t_stat)

if p_val <= significance_level:
    print('Reject Null Hypothesis. There is a statistical evidence that the
    →mean AQI of Michigan is greater than 10')
else:
    print('Fail to reject Null Hypothesis. There not enough statistical
    →evidence that The mean AQI of Michigan is greater than 10')
```

P-value for hypothesis 3: 0.060893005383869395 T-Statistic for hypothesis 3: -1.7395913343286131 Fail to reject Null Hypothesis. There not enough statistical evidence that The mean AQI of Michigan is greater than 10

1.5 Step 4. Results and Evaluation

is the AQI in Los Angeles County was statistically different from the rest of California?** Yes, the results indicated that the AQI in Los Angeles County was in fact different from the rest of California.

Did New York or Ohio have a lower AQI?** Using a 5% significance level, New York has a lower AQI than Ohio based on the results.

Will Michigan be affected by the new policy impacting states with a mean AQI of 10 or greater?** it is unlikely that Michigan would be affected by the new policy.