

Undertanding the cartwheel data set

VANESSA M. CUEVAS ARROYO A01634064

The notebook aims to undertand the content of the cartwheel data set.

Acknowledgments

- Data from <https://www.coursera.org/> from the course "Understanding and Visualizing Data with Python" by University of Michigan

Cartwheel data set

1. A cartwheel

cartwheel1.png

1. The dataset description
 - The dataset used here is an extension from the original cartwheel dataset from cursera
 - Total number of observations: 52
 - Many observations/measurements/recordings of the characteristics/attributes/variables of cartwheel executions
 - Variables: Age, Gender, GenderGroup, Glasses, GlassesGroup, Height, Wingspan, CWDistance, ... (X variables)

Importing and inspecting the data

```
# Define where you are running the code: colab or local
RunInColab = False # (False: no | True: yes)
```

```
# If running in colab:
```

```
if RunInColab:
    # Mount your google drive in google colab
    from google.colab import drive
    drive.mount('/content/drive')
```

```
# Find location
#!pwd
#!ls
#!ls "/content/drive/My Drive/Colab
Notebooks/MachineLearningWithPython/"
```

```
# Define path del proyecto
Ruta = "/content/drive/My Drive/Colab
```

Notebooks/MachineLearningWithPython/"

else:

Define path del proyecto
C:\Users\vanec\Desktop\6to semestre\Analitica\TC1002S\notebooks\Students\A01634064\datasets\cartwheel\carthweel.csv

Ruta = "datasets/cartwheel/cartwheel.csv"

Import the packages that we will be using

import pandas **as** pd

import matplotlib.pyplot **as** plt

Dataset url

Load the dataset

datos = pd.read_csv(Ruta)

Print the dataset

print(datos)

	ID	Age	Gender	GenderGroup	Glasses	GlassesGroup	Height
Wingspan \							
0	1	56.0	F	1	Y	1	62.00
61.0							
1	2	26.0	F	1	Y	1	62.00
60.0							
2	3	33.0	F	1	Y	1	66.00
64.0							
3	4	39.0	F	1	N	0	64.00
63.0							
4	5	27.0	M	2	N	0	73.00
75.0							
5	6	24.0	M	2	N	0	75.00
71.0							
6	7	28.0	M	2	N	0	75.00
76.0							
7	8	22.0	F	1	N	0	65.00
62.0							
8	9	29.0	M	2	Y	1	74.00
73.0							
9	10	33.0	F	1	Y	1	63.00
60.0							
10	11	30.0	M	2	Y	1	69.50
66.0							
11	12	28.0	F	1	Y	1	62.75
58.0							
12	13	25.0	F	1	Y	1	65.00
64.5							

13	14	23.0	F	1	N	0	61.50
57.5							
14	15	31.0	M	2	Y	1	73.00
74.0							
15	16	26.0	M	2	Y	1	71.00
72.0							
16	17	26.0	F	1	N	0	61.50
59.5							
17	18	27.0	M	2	N	0	66.00
66.0							
18	19	23.0	M	2	Y	1	70.00
69.0							
19	20	24.0	F	1	Y	1	68.00
66.0							
20	21	23.0	M	2	Y	1	69.00
67.0							
21	22	29.0	M	2	N	0	71.00
70.0							
22	23	25.0	M	2	N	0	70.00
68.0							
23	24	26.0	M	2	N	0	69.00
71.0							
24	25	23.0	F	1	Y	1	65.00
63.0							
25	26	28.0	M	2	N	0	75.00
76.0							
26	27	24.0	M	2	N	0	78.40
71.0							
27	28	25.0	M	2	Y	1	76.00
73.0							
28	29	32.0	F	1	Y	1	63.00
60.0							
29	30	38.0	F	1	Y	1	61.50
61.0							
30	31	27.0	F	1	Y	1	62.00
60.0							
31	32	33.0	F	1	Y	1	65.30
64.0							
32	33	38.0	F	1	N	0	64.00
63.0							
33	34	27.0	M	2	N	0	77.00
75.0							
34	35	24.0	F	1	N	0	67.80
62.0							
35	36	27.0	M	2	N	0	68.00
66.0							
36	37	25.0	F	1	Y	1	65.00
64.5							
37	38	26.0	F	1	N	0	61.50
59.5							

38	39	31.0	M	2	Y	1	73.00
74.0							
39	40	30.0	M	2	Y	1	69.50
66.0							
40	41	23.0	F	1	N	0	70.40
71.0							
41	42	26.0	M	2	Y	1	73.50
72.0							
42	43	28.0	F	1	Y	1	72.50
72.0							
43	44	26.0	F	1	Y	1	72.00
72.0							
44	45	30.0	F	1	Y	1	66.00
64.0							
45	46	39.0	F	1	N	0	64.00
63.0							
46	47	27.0	M	2	N	0	78.00
75.0							
47	48	24.0	M	2	N	0	79.50
75.0							
48	49	28.0	M	2	N	0	77.80
76.0							
49	50	30.0	F	1	N	0	74.60
NaN							
50	51	NaN	M	2	N	0	71.00
70.0							
51	52	27.0	M	2	N	0	NaN
71.5							

	CWDistance	Complete	CompleteGroup	Score
0	79	Y	1.0	7
1	70	Y	1.0	8
2	85	Y	1.0	7
3	87	Y	1.0	10
4	72	N	0.0	4
5	81	N	0.0	3
6	107	Y	1.0	10
7	98	Y	1.0	9
8	106	N	0.0	5
9	65	Y	1.0	8
10	96	Y	1.0	6
11	79	Y	1.0	10
12	92	Y	1.0	6
13	66	Y	1.0	4
14	72	Y	1.0	9
15	115	Y	1.0	6
16	90	N	0.0	10
17	74	Y	1.0	5
18	64	Y	1.0	3
19	85	Y	1.0	8

20	66	N	0.0	2
21	101	Y	1.0	8
22	82	Y	1.0	4
23	63	Y	1.0	5
24	67	N	0.0	3
25	111	Y	1.0	10
26	92	Y	1.0	7
27	107	Y	1.0	8
28	75	Y	1.0	8
29	78	Y	1.0	7
30	72	Y	1.0	8
31	91	Y	1.0	7
32	86	Y	1.0	10
33	100	Y	1.0	8
34	98	Y	1.0	9
35	74	Y	1.0	5
36	92	Y	1.0	6
37	90	Y	1.0	9
38	72	Y	1.0	9
39	96	Y	1.0	6
40	66	Y	1.0	4
41	115	Y	1.0	6
42	81	Y	1.0	10
43	92	Y	1.0	8
44	85	Y	1.0	7
45	87	Y	1.0	10
46	72	N	0.0	7
47	82	N	0.0	8
48	99	Y	1.0	9
49	71	Y	1.0	9
50	101	Y	NaN	8
51	103	Y	1.0	10

Print the number of rows

```
print("Number of rows: ", len(datos))
```

Number of rows: 52

Print the number of columns

```
print("number of columns", len(datos.columns))
```

number of columns 12

Activity: work with the iris dataset

1. Load the iris.csv file in your computer and understand the dataset
2. How many observations (rows) are in total?
3. How many variables (columns) are in total? What do they represent?

4. How many observations are for each type of flower?
5. What is the type of data for each variable?
6. What are the units of each variable?

#Load the dataset iris.csv

Load the dataset

```
datosIris = pd.read_csv('datasets/iris/iris.csv')
```

Print the number of rows

```
print("Number of rows: ", len(datosIris))
```

Number of rows: 149

Print the number of columns

```
print("number of columns", len(datosIris.columns))
```

```
datosIris.dtypes
```

number of columns 5

```
5.1          float64
3.5          float64
1.4          float64
0.2          float64
Iris-setosa   object
dtype: object
```

```
datosIris.columns = ["PetalWidth", "PetalLength", "SepalWidth",
                     "SepalLength", "Type"]
```

```
datosIris['Type'].value_counts()['Iris-setosa']
```

49

```
datosIris['Type'].value_counts()['Iris-versicolor']
```

50

```
datosIris['Type'].value_counts()['Iris-virginica']
```

50

3.What do they represent?

Las columnas representan las partes de las flores en Sepal y Nepal con su 'weight' and 'Height'

1. How many observations are for each type of flower? 4

2. What is the type of data for each variable? float
3. What are the units of each variable? cm