# Data type constraints

DATA CLEANING IN PYTHON



Adel Nehme
Content Developer @ DataCamp









Side effects of dirty data



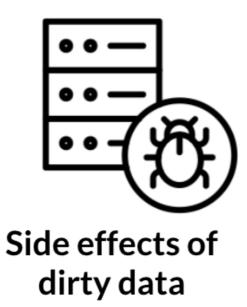


Side effects of dirty data



Clean data







Clean data

Chapter 1 - Common data problems

# Why do we need to clean data?



# Why do we need to clean data?



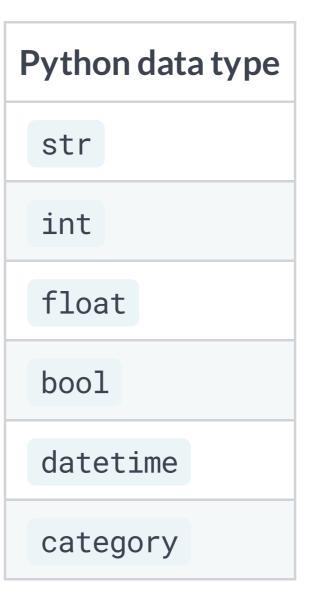
# Why do we need to clean data?



Garbage in Garbage out

# Data type constraints

Datatype	Example
Text data	First name, last name, address
Integers	# Subscribers, # products sold
Decimals	Temperature, \$ exchange rates
Binary	Is married, new customer, yes/no,
Dates	Order dates, ship dates
Categories	Marriage status, gender



# Strings to integers

```
# Import CSV file and output header
sales = pd.read_csv('sales.csv')
sales.head(2)
```

```
SalesOrderID Revenue Quantity
0 43659 23153$ 12
1 43660 1457$ 2
```

```
# Get data types of columns sales.dtypes
```

```
SalesOrderID int64
Revenue object
Quantity int64
dtype: object
```



# String to integers

```
# Get DataFrame information
sales.info()
```



# String to integers

```
# Print sum of all Revenue column
sales['Revenue'].sum()
```

'23153\$1457\$36865\$32474\$472\$27510\$16158\$5694\$6876\$40487\$807\$6893\$9153\$6895\$4216...

```
# Remove $ from Revenue column
sales['Revenue'] = sales['Revenue'].str.strip('$')
sales['Revenue'] = sales['Revenue'].astype(int)
```

```
# Verifiy that Revenue is now an integer
assert sales['Revenue'].dtype == 'int'
```



## The assert statement

```
# This will pass
assert 1+1 == 2

# This will not pass
assert 1+1 == 3
```

```
AssertionError Traceback (most recent call last)

assert 1+1 == 3

AssertionError:
```

# Numeric or categorical?

```
marriage_status
                  3
  = Never married 1 = Married 2 = Separated 3 = Divorced
df['marriage_status'].describe()
      marriage_status
                1.4
mean
                0.20
std
                0.00
min
50%
                1.8 ...
```



# Numeric or categorical?

```
# Convert to categorical
df["marriage_status"] = df["marriage_status"].astype('category')
df.describe()
```

```
marriage_status

count 241

unique 4

top 1

freq 120
```

# Let's practice!

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# Data range constraints

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# Motivation

```
movies.head()
```

```
movie_name avg_rating

0 The Godfather 5

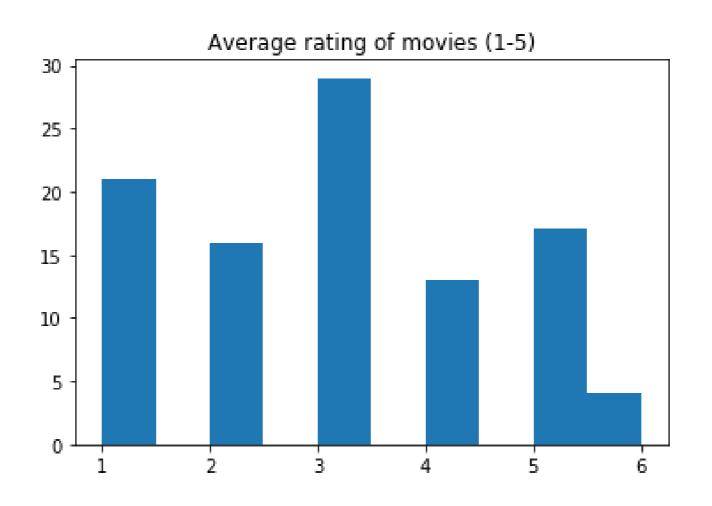
1 Frozen 2 3

2 Shrek 4
...
```



## Motivation

```
import matplotlib.pyplot as plt
plt.hist(movies['avg_rating'])
plt.title('Average rating of movies (1-5)')
```



### Motivation

Can future sign-ups exist?

```
# Import date time
import datetime as dt
today_date = dt.date.today()
user_signups[user_signups['subscription_date'] > dt.date.today()]
```

```
subscription_date
                                                       Country
                   user_name
       01/05/2021
                       Marah
                                                       Nauru
       09/08/2020
                      Joshua
                                                       Austria
                       Heidi
       04/01/2020
                                                       Guinea
       11/10/2020
                                                       Turkmenistan
                        Rina
       11/07/2020
                                                       Marshall Islands
                   Christine
       07/07/2020
                                                       Gabon
                      Ayanna
```

# How to deal with out of range data?

- Dropping data
- Setting custom minimums and maximums
- Treat as missing and impute
- Setting custom value depending on business assumptions

# Movie example

```
import pandas as pd
# Output Movies with rating > 5
movies[movies['avg_rating'] > 5]
```

```
movie_name avg_rating
23 A Beautiful Mind 6
65 La Vita e Bella 6
77 Amelie 6
```

```
# Drop values using filtering
movies = movies[movies['avg_rating'] <= 5]
# Drop values using .drop()
movies.drop(movies[movies['avg_rating'] > 5].index, inplace = True)
# Assert results
assert movies['avg_rating'].max() <= 5</pre>
```



# Movie example

```
# Convert avg_rating > 5 to 5
movies.loc[movies['avg_rating'] > 5, 'avg_rating'] = 5

# Assert statement
assert movies['avg_rating'].max() <= 5</pre>
```

Remember, no output means it passed

# Date range example

```
import datetime as dt
import pandas as pd
# Output data types
user_signups.dtypes
subscription_date
                     object
                     object
user_name
                    object
Country
dtype: object
# Convert to DateTime
user_signups['subscription_date'] = pd.to_datetime(user_signups['subscription_date'])
# Assert that conversion happened
assert user_signups['subscription_date'].dtype == 'datetime64[ns]'
```



# Date range example

```
today_date = dt.date.today()
```

#### Drop the data

```
# Drop values using filtering
user_signups = user_signups[user_signups['subscription_date'] < today_date]
# Drop values using .drop()
user_signups.drop(user_signups[user_signups['subscription_date'] > today_date].index, inplace = True)
```

### Hardcode dates with upper limit

```
# Drop values using filtering
user_signups.loc[user_signups['subscription_date'] > today_date, 'subscription_date'] = today_date
# Assert is true
assert data.subscription_date.max().date() <= today_date</pre>
```

# Let's practice!

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# Uniqueness constraints

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Adel Nehme
Content Developer @ DataCamp



# What are duplicate values?

#### All columns have the same values

first_name	last_name	address	height	weight
Justin	Saddlemyer	Boulevard du Jardin Botanique 3, Bruxelles	193 cm	87 kg
Justin	Saddlemyer	Boulevard du Jardin Botanique 3, Bruxelles	193 cm	87 kg

# What are duplicate values?

#### Most columns have the same values

first_name	last_name	address	height	weight
Justin	Saddlemyer	Boulevard du Jardin Botanique 3, Bruxelles	193 cm	87 kg
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# Why do they happen?



Data Entry & Human Error

# Why do they happen?



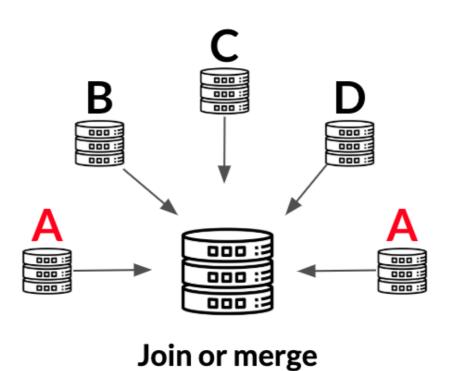
Data Entry & Human Error



Bugs and design errors

# Why do they happen?





**Errors** 



# How to find duplicate values?

```
# Print the header
height_weight.head()
```

ı		first_name	last_name	address	height	weight
ı	0	Lane	Reese	534-1559 Nam St.	181	64
ı	1	Ivor	Pierce	102-3364 Non Road	168	66
ı	2	Roary	Gibson	P.O. Box 344, 7785 Nisi Ave	191	99
ı	3	Shannon	Little	691-2550 Consectetuer Street	185	65
	4	Abdul	Fry	4565 Risus St.	169	65



# How to find duplicate values?

```
# Get duplicates across all columns
duplicates = height_weight.duplicated()
print(duplicates)
```

```
1 False
... ...
22 True
23 False
... ...
```

# How to find duplicate values?

```
# Get duplicate rows
duplicates = height_weight.duplicated()
height_weight[duplicates]
```

f:	irst_name [	last_name	address	height	weight
100	Mary	Colon	4674 Ut Rd.	179	75
101	Ivor	Pierce	102-3364 Non Road	168	88
102	Cole	Palmer	8366 At, Street	178	91
103	Desirae	Shannon	P.O. Box 643, 5251 Consectetuer, Rd.	196	83

```
The .duplicated() method

subset : List of column names to check for duplication.

keep : Whether to keep first( 'first'), last( 'last') or all( False ) duplicate values.

# Column names to check for duplication
column_names = ['first_name', 'last_name', 'address']
duplicates = height_weight.duplicated(subset = column_names, keep = False)
```

# Output duplicate values
height\_weight[duplicates]

	first_name	last_name		address	height	weight
1	Ivor	Pierce		102-3364 Non Road	168	66
22	Cole	Palmer		8366 At, Street	178	91
28	B Desirae	Shannon	P.O. Box 643, 525	1 Consectetuer, Rd.	195	83
37	7 Mary	Colon		4674 Ut Rd.	179	75
18	Mary	Colon		4674 Ut Rd.	179	75
18	1 Ivor	Pierce		102-3364 Non Road	168	88
18	Cole	Palmer		8366 At, Street	178	91
16	Desirae	Shannon	P.O. Box 643, 525	1 Consectetuer, Rd.	196	83



```
# Output duplicate values
height_weight[duplicates].sort_values(by = 'first_name')
```

	first_name ]	last_name					а	ddress	height	weight	
22	Cole	Palmer					8366 At,	Street	178	91	
102	Cole	Palmer					8366 At,	Street	178	91	
28	Desirae	Shannon	P.0.	Box	643,	5251	Consectetue	r, Rd.	195	83	
103	Desirae	Shannon	P.0.	Box	643,	5251	Consectetue	r, Rd.	196	83	
1	Ivor	Pierce					102-3364 No	n Road	168	66	
101	Ivor	Pierce					102-3364 No	n Road	168	88	
37	Mary	Colon					4674	Ut Rd.	179	75	
100	Mary	Colon					4674	Ut Rd.	179	75	



```
# Output duplicate values
height_weight[duplicates].sort_values(by = 'first_name')
```

	first_name	last_name		address	height	weight
22	Cole	Palmer		8366 At, Street	178	91
102	Cole	Palmer		8366 At, Street	178	91
28	Desirae	Shannon	P.O. Box 643, 5	251 Consectetuer, Rd.	195	83
103	Desirae	Shannon	P.O. Box 643, 5	251 Consectetuer, Rd.	196	83
1	Ivor	Pierce		102-3364 Non Road	168	66
101	Ivor	Pierce		102-3364 Non Road	168	88
37	Mary	Colon		4674 Ut Rd.	179	75
100	Mary	Colon		4674 Ut Rd.	179	75



```
# Output duplicate values
height_weight[duplicates].sort_values(by = 'first_name')
```

	first_name	last_name	address	height	weight
22	Cole	Palmer	8366 At, Street	178	91
102	Cole	Palmer	8366 At, Street	178	91
28	Desirae	Shannon	P.O. Box 643, 5251 Consectetuer, Rd.	195	83
103	Desirae	Shannon	P.O. Box 643, 5251 Consectetuer, Rd.	196	83
1	Ivor	Pierce	102-3364 Non Road	168	66
101	Ivor	Pierce	102-3364 Non Road	168	88
37	Mary	Colon	4674 Ut Rd.	179	75
100	Mary	Colon	4674 Ut Rd.	179	75



```
# Output duplicate values
height_weight[duplicates].sort_values(by = 'first_name')
```

	first_name	last_name		address	height	weight
22	Cole	Palmer		8366 At, Street	178	91
102	Cole	Palmer		8366 At, Street	178	91
28	Desirae	Shannon	P.O. Box 643, 52	251 Consectetuer, Rd.	195	83
103	Desirae	Shannon	P.O. Box 643, 52	251 Consectetuer, Rd.	196	83
1	Ivor	Pierce		102-3364 Non Road	168	66
101	Ivor	Pierce		102-3364 Non Road	168	88
37	Mary	Colon		4674 Ut Rd.	179	75
100	Mary	Colon		4674 Ut Rd.	179	75



```
The .drop_duplicates() method
 subset: List of column names to check for duplication.
 keep: Whether to keep first ('first'), last ('last') or all (False) duplicate values.
 inplace: Drop duplicated rows directly inside DataFrame without creating new object ( True ).
 # Drop duplicates
 height_weight.drop_duplicates(inplace = True)
```

```
# Output duplicate values
column_names = ['first_name','last_name','address']
duplicates = height_weight.duplicated(subset = column_names, keep = False)
height_weight[duplicates].sort_values(by = 'first_name')
```

ı		first_name	last_name					address	height	weight	
ı	28	Desirae	Shannon	P.0.	Box	643,	5251	Consectetuer, Rd.	195	83	
ı	103	Desirae	Shannon	P.O.	Box	643,	5251	Consectetuer, Rd.	196	83	
ı	1	Ivor	Pierce					102-3364 Non Road	168	66	
ı	101	Ivor	Pierce					102-3364 Non Road	168	88	



```
# Output duplicate values
column_names = ['first_name','last_name','address']
duplicates = height_weight.duplicated(subset = column_names, keep = False)
height_weight[duplicates].sort_values(by = 'first_name')
```

	first_name	last_name		address	height	weight
28	Desirae	Shannon	P.O. Box 643,	5251 Consectetuer, Rd.	195	83
103	Desirae	Shannon	P.O. Box 643,	5251 Consectetuer, Rd.	196	83
1	Ivor	Pierce		102-3364 Non Road	168	66
101	Ivor	Pierce		102-3364 Non Road	168	88

The .groupby() and .agg() methods

```
# Group by column names and produce statistical summaries
column_names = ['first_name','last_name','address']
summaries = {'height': 'max', 'weight': 'mean'}
height_weight = height_weight.groupby(by = column_names).agg(summaries).reset_index()

# Make sure aggregation is done
duplicates = height_weight.duplicated(subset = column_names, keep = False)
height_weight[duplicates].sort_values(by = 'first_name')
```

```
first_name last_name address height weight
```



# Let's practice!

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