

## Absolute vs. Relative Frequency

By default, seaborn's countplot function will summarize and plot the data in terms of **absolute frequency**, or pure counts. In certain cases, you might want to understand the distribution of data or want to compare levels in terms of the proportions of the whole. In this case, you will want to plot the data in terms of **relative frequency**, where the height indicates the proportion of data taking each level, rather than the absolute count.

One method of plotting the data in terms of relative frequency on a bar chart is to just relabel the count's axis in terms of proportions. The underlying data will be the same, it will simply be the scale of the axis ticks that will be changed.

**Example 1. Demonstrate data wrangling, and plot a horizontal bar chart.**

**Example 1 - Step 1. Make the necessary import**

```
import numpy as np
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sb
```

```
%matplotlib inline
```

```
# Read the data from a CSV file
```

```
pokemon = pd.read_csv('pokemon.csv')
```

```
print(pokemon.shape)
```

```
pokemon.head(10)
```

	id	species	generation_id	height	weight	base_experience	type_1	type_2	hp	attack	defense	speed	special-attack	special-defense
0	1	bulbasaur	1	0.7	6.9	64	grass	poison	45	49	49	45	65	65
1	2	ivysaur	1	1.0	13.0	142	grass	poison	60	62	63	60	80	80
2	3	venusaur	1	2.0	100.0	236	grass	poison	80	82	83	80	100	100
3	4	charmander	1	0.6	8.5	62	fire	NaN	39	52	43	65	60	50
4	5	charmeleon	1	1.1	19.0	142	fire	NaN	58	64	58	80	80	65
5	6	charizard	1	1.7	90.5	240	fire	flying	78	84	78	100	109	85
6	7	squirtle	1	0.5	9.0	63	water	NaN	44	48	65	43	50	64
7	8	wartortle	1	1.0	22.5	142	water	NaN	59	63	80	58	65	80
8	9	blastoise	1	1.6	85.5	239	water	NaN	79	83	100	78	85	105
9	10	caterpie	1	0.3	2.9	39	bug	NaN	45	30	35	45	20	20

Last time we created the bar chart of pokemon by their type\_1. Let's club the rows of both type\_1 and type\_2, so that the resulting dataframe has **new** column, type\_level.

This operation will double the number of rows in pokemon from 807 to 1614.

## Data Wrangling Step

We will use the `pandas.DataFrame.melt()` method to unpivot a DataFrame from wide to long format, optionally leaving identifiers set. The syntax is:

```
DataFrame.melt(id_vars, value_vars, var_name, value_name, col_level, ignore_index)
```

It is essential to understand the parameters involved:

1. `id_vars` - It is a tuple representing the column(s) to use as identifier variables.
2. `value_vars` - It is tuple representing the column(s) to unpivot (remove, out of place).
3. `var_name` - It is a name of the **new** column.
4. `value_name` - It is a name to use for the 'value' of the columns that are unpivoted.

Refer [here](#) for more details on the parameters.

The function below will do the following in the pokemon dataframe *out of place*:

1. Select the 'id', and 'species' columns from pokemon.
2. Remove the 'type\_1', 'type\_2' columns from pokemon
3. Add a new column 'type\_level' that can have a value either 'type\_1' or 'type\_2'
4. Add another column 'type' that will contain the actual value contained in the 'type\_1', 'type\_2' columns. For example, the first row in the pokemon dataframe having id=1 and species=bulbasaur will now occur twice in the resulting dataframe after the melt() operation. The first occurrence will have type=grass, whereas, the second occurrence will have type=poison.

### Example 1 - Step 2. Data wrangling to reshape the pokemon dataframe

```
pkmn_types = pokemon.melt(id_vars=['id', 'species'],  
                           value_vars=['type_1', 'type_2'],  
                           var_name='type_level',  
                           value_name='type')  
pkmn_types.head(10)  
#pkmn_types.shape
```

	<b>id</b>	<b>species</b>	<b>type_level</b>	<b>type</b>
<b>0</b>	1	bulbasaur	type_1	grass
<b>1</b>	2	ivysaur	type_1	grass
<b>2</b>	3	venusaur	type_1	grass
<b>3</b>	4	charmander	type_1	fire
<b>4</b>	5	charmeleon	type_1	fire
<b>5</b>	6	charizard	type_1	fire
<b>6</b>	7	squirtle	type_1	water
<b>7</b>	8	wartortle	type_1	water
<b>8</b>	9	blastoise	type_1	water
<b>9</b>	10	caterpie	type_1	bug

**Example 1 - Step 3. Find the frequency of unique values in the type column**

```
# Count the frequency of unique values in the `type` column of pkmn_types dataframe.
```

```
# By default, returns the decreasing order of the frequency.
```

```
type_counts = pkmn_types['type'].value_counts()
```

```
type_counts
```

water	131
normal	109
flying	98
grass	97
psychic	82
bug	77
poison	66
ground	64
fire	64
rock	60
fighting	54
electric	48
fairy	47
steel	47
dark	46
dragon	45
ghost	43
ice	34

Name: type, dtype: int64

```
# Get the unique values of the `type` column, in the decreasing order of the frequency.
```

```
type_order = type_counts.index
```

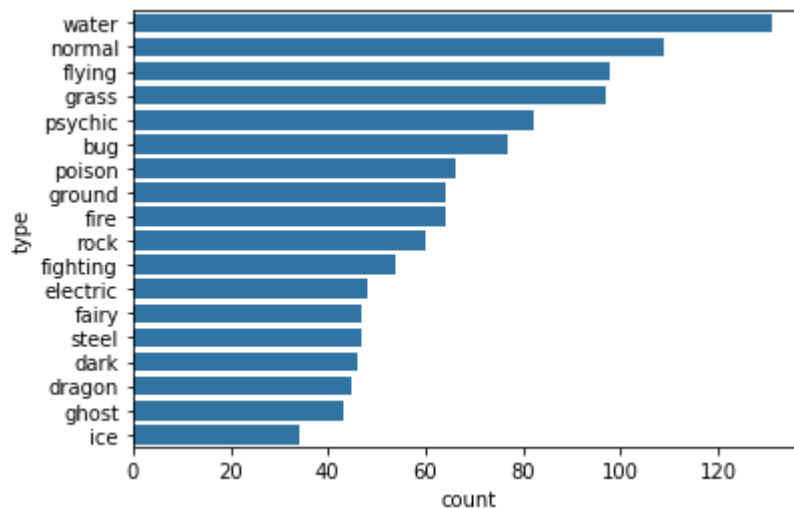
```
type_order
```

```
Index(['water', 'normal', 'flying', 'grass', 'psychic', 'bug', 'poison', 'ground', 'fire', 'rock', 'fighting',  
'electric', 'fairy', 'steel', 'dark', 'dragon', 'ghost', 'ice'], dtype='object')
```

#### Example 1 - Step 4. Plot the horizontal bar charts

```
base_color = sb.color_palette()[0]
```

```
sb.countplot(data=pkmn_types, y='type', color=base_color, order=type_order);
```



#### Example 2. Plot a bar chart having the proportions, instead of the actual count, on one of the axes.

##### Example 2 - Step 1. Find the maximum proportion of bar

```
# Returns the sum of all not-null values in `type` column
```

```
n_pokemon = pkmn_types['type'].value_counts().sum()
```

```
# Return the highest frequency in the `type` column
```

```
max_type_count = type_counts[0]
```

```
# Return the maximum proportion, or in other words,
```

```
# compute the length of the longest bar in terms of the proportion
```

```
max_prop = max_type_count / n_pokemon
```

```
print(max_prop)
```

0.1623296158612144

### Example 2 - Step 2. Create an array of evenly spaced proportioned values

```
# Use numpy.arange() function to produce a set of evenly spaced proportioned values
```

```
# between 0 and max_prop, with a step size 2\%
```

```
tick_props = np.arange(0, max_prop, 0.02)
```

```
tick_props
```

```
array([0. , 0.02, 0.04, 0.06, 0.08, 0.1 , 0.12, 0.14, 0.16])
```

We need x-tick labels that must be evenly spaced on the x-axis. For this purpose, we must have a list of labels ready with us, before using it with `plt.xticks()` function.

### Example 2 - Step 3. Create a list of String values that can be used as tick labels.

```
# Use a list comprehension to create tick_names that we will apply to the tick labels.
```

```
# Pick each element `v` from the `tick_props`, and convert it into a formatted string.
```

```
# `{:0.2f}` denotes that before formatting, we 2 digits of precision and `f` is used to represent floating point number.
```

```
# Refer [here](https://docs.python.org/2/library/string.html#format-string-syntax) for more details
```

```
tick_names = ['{:0.2f}'.format(v) for v in tick_props]
```

```
tick_names
```

```
['0.00', '0.02', '0.04', '0.06', '0.08', '0.10', '0.12', '0.14', '0.16']
```

The [xticks](#) and [yticks](#) functions aren't only about rotating the tick labels. You can also get and set their locations and labels as well. The first argument takes the tick locations: in this case, the tick proportions multiplied back to be on the scale of counts. The second argument takes the tick names: in this case, the tick proportions formatted as strings to two decimal places.

I've also added a [ylabel](#) call to make it clear that we're no longer working with straight counts.

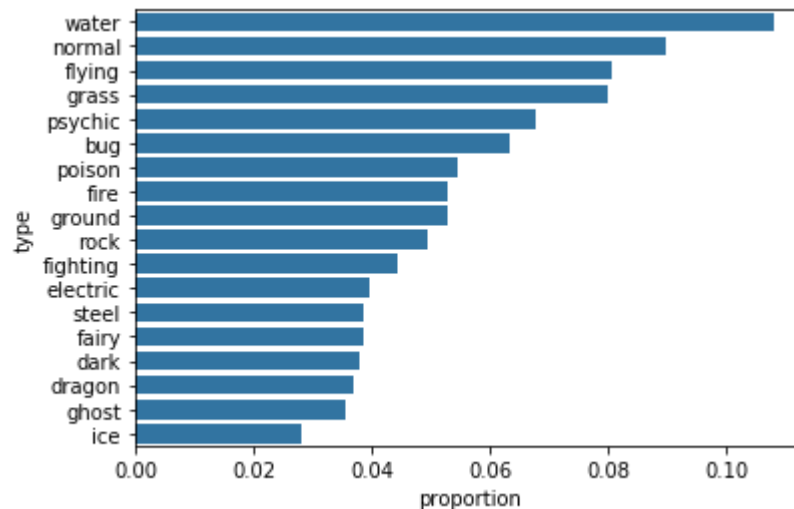
### Example 2 - Step 4. Plot the bar chart, with new x-tick labels

```
sb.countplot(data=pkmn_types, y='type', color=base_color, order=type_order);
```

```
# Change the tick locations and labels
```

```
plt.xticks(tick_props * n_pokemon, tick_names)
```

```
plt.xlabel('proportion');
```



### Additional Variation

Rather than plotting the data on a relative frequency scale, you might use text annotations to label the frequencies on bars instead. This requires writing a loop over the tick locations and labels and adding one text element for each bar.

#### Example 3. Print the text (proportion) on the bars of a horizontal plot.

```
# Considering the same chart from the Example 1 above, print the text (proportion) on the bars
```

```
base_color = sb.color_palette()[0]
```

```
sb.countplot(data=pkmn_types, y='type', color=base_color, order=type_order);
```

```
# Logic to print the proportion text on the bars
```

```
for i in range(type_counts.shape[0]):
```

```
    # Remember, type_counts contains the frequency of unique values in the `type` column in decreasing order.
```

```
    count = type_counts[i]
```

```
    # Convert count into a percentage, and then into string
```

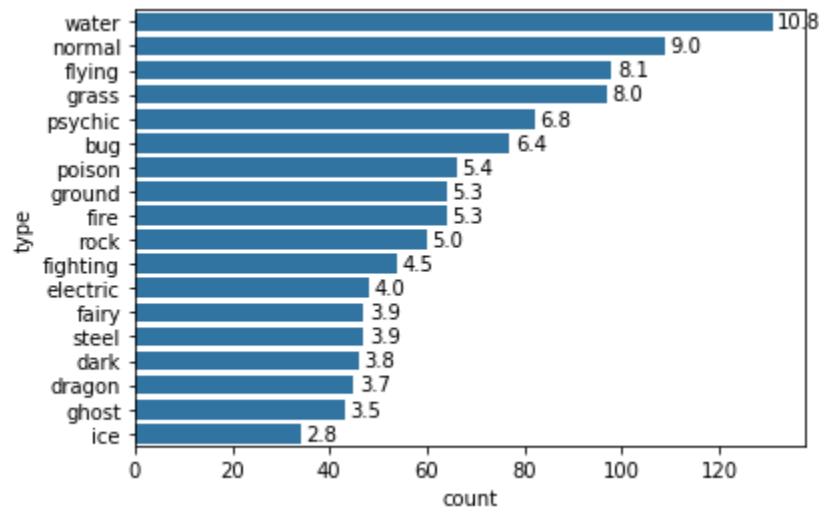
```
    pct_string = '{:0.1f}'.format(100*count/n_pokemon)
```

```
    # Print the string value on the bar.
```

```
    # Read more about the arguments of text() function
```

```
[here](https://matplotlib.org/3.1.1/api/_as_gen/matplotlib.pyplot.text.html)
```

```
    plt.text(count+1, i, pct_string, va='center')
```



**Example 4. Print the text (proportion) below the bars of a Vertical plot.**

# Considering the same chart from the Example 1 above, print the text (proportion) BELOW the bars

```
base_color = sb.color_palette()[0]
```

```
sb.countplot(data=pkmn_types, x='type', color=base_color, order=type_order);
```

# Recalculating the type\_counts just to have clarity.

```
type_counts = pkmn_types['type'].value_counts()
```

# get the current tick locations and labels

```
locs, labels = plt.xticks(rotation=90)
```

# loop through each pair of locations and labels

```
for loc, label in zip(locs, labels):
```

```
    # get the text property for the label to get the correct count
```

```
    count = type_counts[label.get_text()]
```

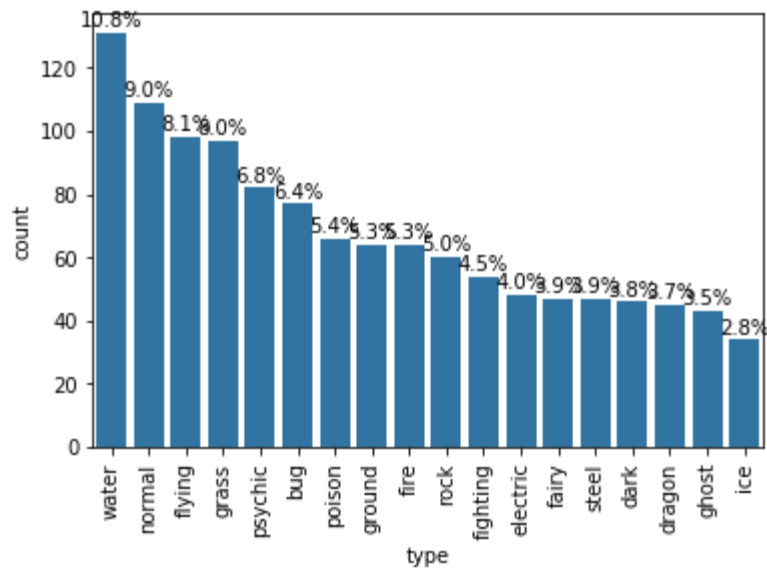
```
    pct_string = '{:0.1f}%'.format(100*count/n_pokemon)
```

```
    # print the annotation just below the top of the bar
```



```
plt.text(loc, count+2, pct_string, ha = 'center', color = 'black')
```

I use the `.get_text()` method to obtain the category name, so I can get the count of each category level. At the end, I use the [text](#) function to print each percentage, with the x-position, y-position, and string as the three main parameters to the function.



**Tip** - Is the text on the bars not readable clearly? Consider changing the size of the plot by using the following:

```
from matplotlib import rcParams
```

```
# Specify the figure size in inches, for both X, and Y axes
```

```
rcParams['figure.figsize'] = 12,4
```

