

Analysis Of Land Used in Pakistan

Agricultural Vs Other Land

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In [1]: import numpy as np
import pandas as pd
from numpy import *
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: land = pd.read_csv("land_used.csv")
```

```
In [3]: land.head(3)
```

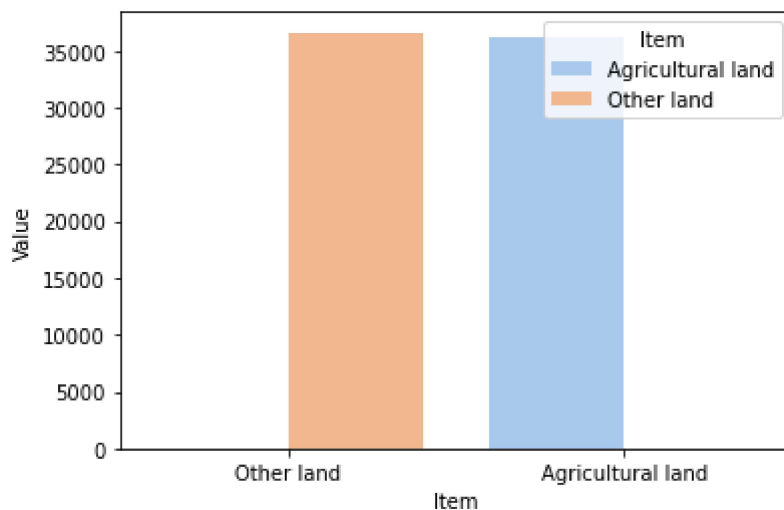
Out[3]:

	Domain Code	Domain	Area Code	Area	Element Code	Element	Item Code	Item	Year Code	Year	Unit	Value
0	RL	Land Use	165	Pakistan	5110	Area	6610	Agricultural land	1961	1961	1000 ha	35730.0
1	RL	Land Use	165	Pakistan	5110	Area	6610	Agricultural land	1962	1962	1000 ha	35840.0
2	RL	Land Use	165	Pakistan	5110	Area	6610	Agricultural land	1963	1963	1000 ha	35880.0

```
In [4]: #land.groupby(["Item",])

sns.barplot(x="Item",y="Value",data=land,hue = "Item", order = ["Other land","Agricultural land"],
            ci=None, saturation =0.8)
```

Out[4]: <AxesSubplot:xlabel='Item', ylabel='Value'>



```
In [5]: # now grouping values
land.groupby(["Item", "Year"]).mean()
```

```
Out[5]:
```

		Area Code	Element Code	Item Code	Year Code	Value
	Item	Year				
Agricultural land	1961		165.0	5110.0	6610.0	1961.0 35730.00
			165.0	5110.0	6610.0	1962.0 35840.00
			165.0	5110.0	6610.0	1963.0 35880.00
			165.0	5110.0	6610.0	1964.0 36355.00
			165.0	5110.0	6610.0	1965.0 37235.00

Other land	2015		165.0	5110.0	6670.0	2015.0 36953.40
			165.0	5110.0	6670.0	2016.0 36425.76
			165.0	5110.0	6670.0	2017.0 36235.08
			165.0	5110.0	6670.0	2018.0 36979.42
			165.0	5110.0	6670.0	2019.0 37020.76

89 rows × 5 columns

```
In [6]: # now Lets see for 2015 after stats
land[land["Year"]>2015].groupby(["Item", "Year"]).mean().head(5)
```

```
Out[6]:
```

		Area Code	Element Code	Item Code	Year Code	Value
	Item	Year				
Agricultural land	2016		165.0	5110.0	6610.0	2016.0 36794.00
			165.0	5110.0	6610.0	2017.0 37003.00
			165.0	5110.0	6610.0	2018.0 36300.00
			165.0	5110.0	6610.0	2019.0 36300.00
Other land	2016		165.0	5110.0	6670.0	2016.0 36425.76

```
In [ ]:
```