## **Craft-Cans Case Study**

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# R Markdown Dummy

First

### **About the Data**

#### **Source**

- The tables contain a list of 2410 US craft beers and 510 US breweries.
- The beer data corresponds to beers available in cans
- This data was traced from CraftCans.com.

### **About the Data**

#### View the data

```
## Load Data ##
#getwd()
setwd("C:/Users/acasi/Downloads")
BeersData <- read.csv("Beers.csv")
BreweriesData <- read.csv("Breweries.csv")
str(BeersData)
## 'data.frame': 2410 obs. of 7 variables:
## $ Name : Factor w/ 2305 levels "#001 Golden Amber Lag</pre>
```

## \$ IBU : int NA ...

\$ Beer ID : int 1436 2265 2264 2263 2262 2261 2260 2259

: num 0.05 0.066 0.071 0.09 0.075 0.077 0.049

\$ ABV

##

##

# Munging

#### **Basic alterations**

- Column names are altered for clarity and to minimize merging issues
- The beers and breweries are linked by a numeric ID, which is used to merge the two tables

"ABV"

"IBU"

'Brew\_ID'

## [1] "Name"

```
## Prepare for merging ##
names(BeersData)
```

```
## [6] "Style" "Ounces"
```

"Beer ID"

```
names(BeersData)[5] = "Brew_ID"
names(BeersData)[1] = "Beer.name"
names(BreweriesData)[2] = "Brewery.name"
```

# Munging

### View the final data

-first 6 rows

```
library(knitr)
#library(kableExtra)
kable(head(AllBeer, 6))
```

Brew_ID	Beer.name	Beer_ID	ABV	IBU	Style
1	Get Together	2692	0.045	50	American IPA
1	Maggie's Leap	2691	0.049	26	Milk / Sweet Stout
1	Wall's End	2690	0.048	19	English Brown Ale
1	Pumpion	2689	0.060	38	Pumpkin Ale
1	Stronghold	2688	0.060	25	American Porter
1	Parapet ESB	2687	0.056	47	Extra Special / Strong

# Munging

#### View the final data

last 6 rows

```
library(knitr)
kable(tail(AllBeer, 6))
```

	Brew_ID	Beer.name	Beer_ID	ABV	IBU	Sty
2405	556	Pilsner Ukiah	98	0.055	NA	Ge
2406	557	Heinnieweisse Weissebier	52	0.049	NA	He
2407	557	Snapperhead IPA	51	0.068	NA	An
2408	557	Moo Thunder Stout	50	0.049	NA	Mi
2409	557	Porkslap Pale Ale	49	0.043	NA	An
2410	558	Urban Wilderness Pale Ale	30	0.049	NA	En
						_

## **Data Integrity**

### **Missing Vlaues**

Check the missing values number and indicators

```
## Check for missing values ##
apply(apply(AllBeer, 2, is.na), 2, sum)
```

```
## Brew_ID Beer.name Beer_ID ABV
## 0 0 0 0 62
## Style Ounces Brewery.name City
## 0 0 0 0
```

```
t=sapply(AllBeer, function(y) sum(length(which(is.na(y)))))
summary(t)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
```

## **Data Integrity**

#### **Notes and Observations**

- Only ABV and IBU have missing values
- Almost 50% of IBU values are missing -This certainly affect any analysis that hinges on IBU values
- Spot checks on the data suggest that character variables use blanks for missing values
- This needs to be verified

## **Data Integrity**

## Double Check ##

### **Double Check Missing Vlaues**

## Look at balnk Strings ##

Check for blanks instead of NA's

```
apply(AllBeer, 2, function(y) sum(y == ""))

## Brew_ID Beer.name Beer_ID ABV
## 0 0 0 NA
## Style Ounces Brewery.name City
## 5 0 0 0
```

'Stye', which represents style of beer, has 5 additional missing values

### **External Data**

- From R
- R contains additional region and division data per state in the package 'Datasets'
- This data is from a 1977 report from the chamber of commerce

## **External Data**

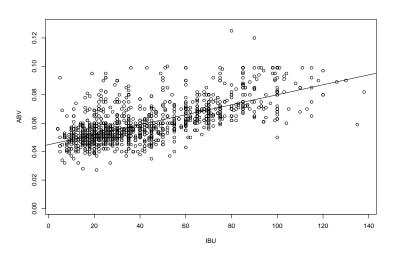
- From census
- This is a 2016 geography table

### Tail2

### -experiment with plotting last 6 rows

Brew_ID	Beer.name	Beer_ID	ABV	IBU	Style
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# **Simple Plot**



## **Fancier Plot**

