FP2

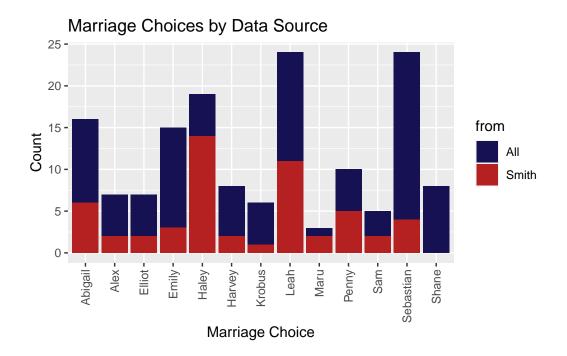
Loading and Cleaning the Data

```
# reading in csvs and creating df
library(tidyverse)

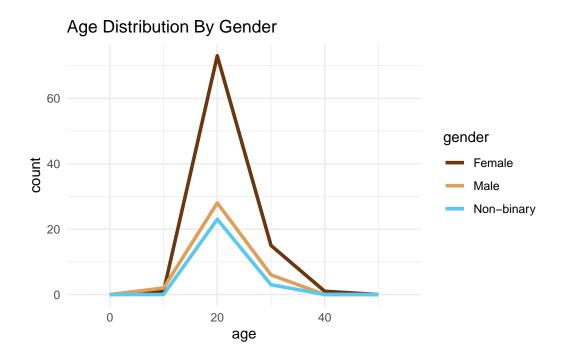
stardew_all <- read.csv("stardew_all.csv") |> select(!Timestamp) |> mutate(from = "All")
stardew_smith <- read.csv("stardew_smith.csv") |> select(!Timestamp) |> mutate(from = "Smi
stardew_data <- rbind(stardew_all, stardew_smith)

stardew_data <- stardew_data |> filter(gender != "Other")
```

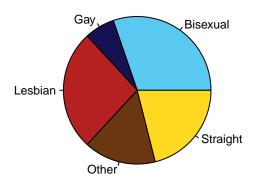
Distribution of The Data



Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0. i Please use `linewidth` instead.



```
sexuality_freq <- table(stardew_data$sexuality)
pie(sexuality_freq, cex=.75, radius = 1, col = c('#59C9F1', '#151152', '#B52121', '#6B3710')</pre>
```



Splitting Up Data for Cross Validation

```
# separating the data into train and test data
set.seed(1) # set seed

indices <- sample(seq(152), 106) # find 108 random indices

train <- stardew_data[indices,] # use those indices for training data

test_data <- stardew_data[-indices,] # use all the other ones for testing
test_marriage <- stardew_data$marriage_candidate[-indices] # get true marriage candidates</pre>
```

Naive Bayes Model

```
# Naive Bayes
  library(e1071)
  train <- indices
  nb.fit <- naiveBayes(marriage_candidate ~ gender + sexuality + age, data = stardew_data, s
  nb.fit
Naive Bayes Classifier for Discrete Predictors
Call:
naiveBayes.default(x = X, y = Y, laplace = laplace)
A-priori probabilities:
                   Alex
                             Elliot
    Abigail
                                          Emily
                                                       Haley
                                                                  Harvey
0.075471698\ 0.037735849\ 0.066037736\ 0.084905660\ 0.113207547\ 0.047169811
                   Leah
                               Maru
                                          Penny
                                                         Sam
                                                               Sebastian
0.047169811 0.216981132 0.009433962 0.056603774 0.047169811 0.141509434
      Shane
0.056603774
Conditional probabilities:
           gender
γ
                             Male Non-binary
                Female
  Abigail 0.37500000 0.62500000 0.00000000
           0.25000000 0.75000000 0.00000000
  Alex
```

```
Elliot
           0.85714286 0.00000000 0.14285714
           0.4444444 0.2222222 0.33333333
 Emily
           0.58333333 0.16666667 0.25000000
 Haley
           1.00000000 0.00000000 0.00000000
 Harvey
           0.60000000 0.00000000 0.40000000
 Krobus
 Leah
           0.60869565 0.26086957 0.13043478
 Maru
           1.00000000 0.00000000 0.00000000
 Penny
           0.66666667 0.33333333 0.00000000
           0.4000000 0.60000000 0.00000000
 Sam
 Sebastian 0.66666667 0.06666667 0.26666667
           0.3333333 0.33333333 0.33333333
 Shane
          sexuality
Y
             Bisexual
                            Gay
                                   Lesbian
                                               Other
                                                       Straight
           0.25000000 0.00000000 0.12500000 0.00000000 0.62500000
 Abigail
 Alex
           0.25000000 0.25000000 0.25000000 0.00000000 0.25000000
 Elliot
           0.28571429 0.14285714 0.00000000 0.14285714 0.42857143
 Emily
           0.11111111 0.00000000 0.55555556 0.11111111 0.22222222
 Haley
           0.25000000 0.08333333 0.58333333 0.08333333 0.00000000
 Harvey
           0.60000000 0.00000000 0.00000000 0.20000000 0.20000000
           0.00000000 0.00000000 0.20000000 0.80000000 0.00000000
 Krobus
           0.08695652 0.00000000 0.56521739 0.08695652 0.26086957
 Leah
 Maru
           0.0000000 0.0000000 1.0000000 0.0000000 0.0000000
           0.16666667 0.00000000 0.33333333 0.16666667 0.33333333
 Penny
 Sam
           Sebastian 0.40000000 0.06666667 0.06666667 0.33333333 0.13333333
           Shane
          age
Y
               [,1]
                        [,2]
 Abigail
           22.00000 4.000000
 Alex
           20.12500 1.436141
 Elliot
           21.28571 2.288689
           21.22222 3.032234
 Emily
 Haley
           20.16667 1.466804
 Harvey
           25.20000 5.069517
 Krobus
           25.60000 4.159327
 Leah
           21.65217 4.923124
 Maru
           20.00000
                         NA
 Penny
           21.16667 3.868678
           20.60000 2.701851
 Sam
 Sebastian 23.06667 4.431489
```

Shane

22.83333 5.231316

```
nb.class <- predict(nb.fit, test_data)
table(nb.class, test_data$marriage_candidate)</pre>
```

nb.class	Abigail	Alex	Emily	Haley	Harvey	Krobus	Leah	Maru	Penny	Sebastian
Abigail	2	0	1	0	0	0	0	0	1	0
Alex	0	0	0	0	1	0	0	0	0	0
Elliot	0	0	0	0	0	0	0	0	0	2
Emily	0	0	0	0	0	0	0	0	0	0
Haley	3	0	0	5	1	0	0	1	1	2
Harvey	0	0	0	0	0	0	0	0	0	1
Krobus	0	1	1	0	0	0	0	0	0	0
Leah	0	0	1	1	0	0	1	0	0	0
Maru	0	0	0	0	0	0	0	0	0	0
Penny	0	0	0	0	0	0	0	0	0	0
Sam	1	2	0	0	0	0	0	0	0	0
Sebastian	2	0	2	1	1	1	0	1	2	4
Shane	0	0	1	0	0	0	0	0	0	0

nb.class	Shane
Abigail	1
Alex	0
Elliot	0
Emily	0
Haley	0
Harvey	0
Krobus	0
Leah	0
Maru	0
Penny	0
Sam	0
Sebastian	1
Shane	0

```
mean(nb.class == test_data$marriage_candidate)
```

[1] 0.2608696