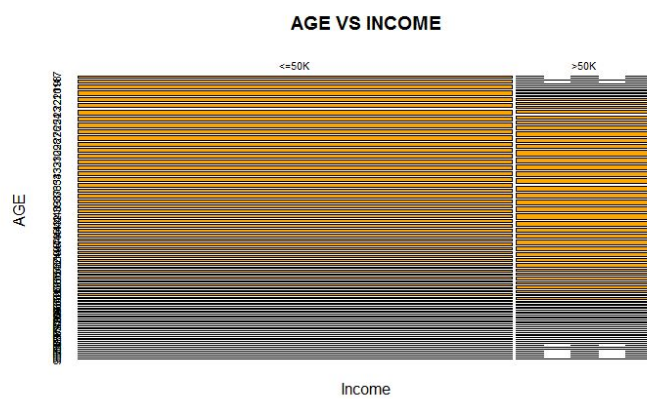
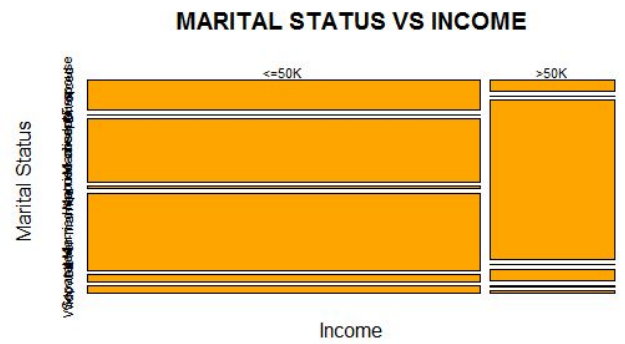
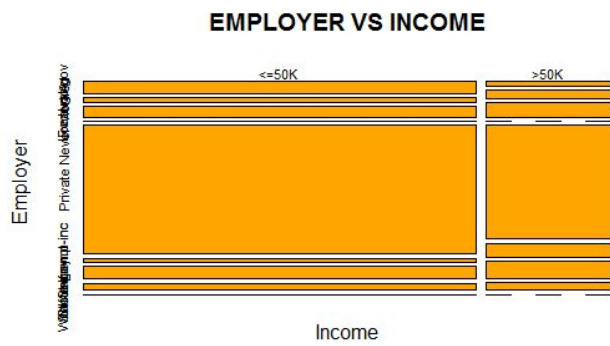
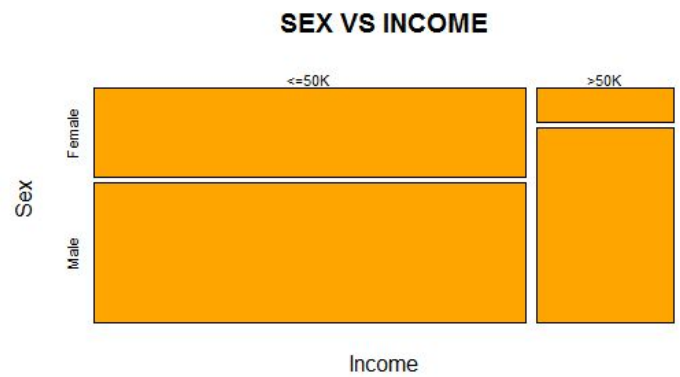
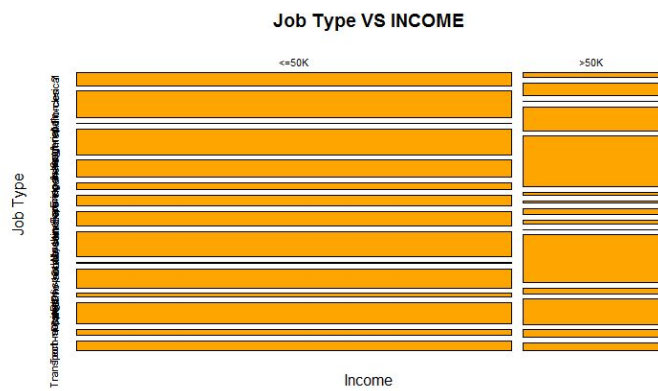


Problem 1:

- A) There are 32,561 records. The function in R that you can use to get this information is  
`dim(salary_class)`
- B) Measurements of all the variables:  
AGE: Nominal; discrete  
Employer: Categorical - Nominal  
Degree: Ordinal-discrete  
MStatus: Nominal  
Jobtype: nominal  
Sex: binary  
C-Gain: Interval - Integer  
C-Loss: Interval - Integer  
Hours: Interval - Integer  
Country: Nominal  
Income: Binary

You can find these measurements by using the `str(salary_class)` function in R.

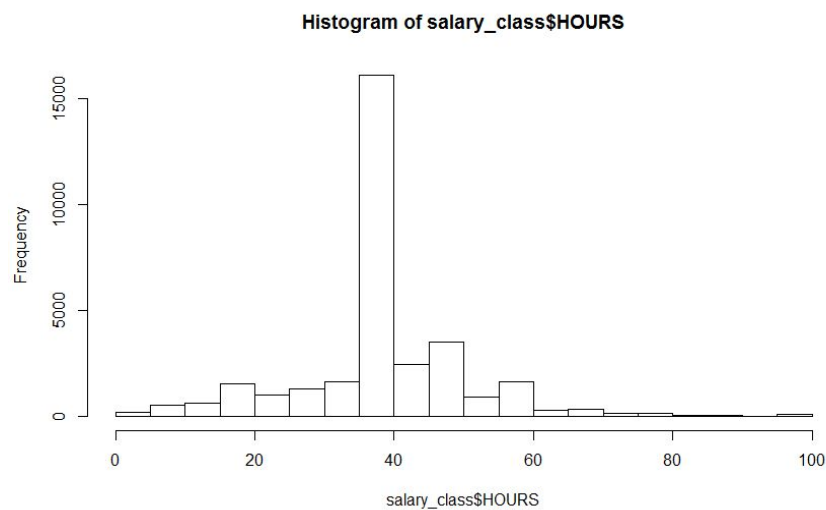
- C) Age Group:
- a) Mean: 38.58
  - b) Median: 37
  - c) Variance: 186.06
  - d) Standard Deviation: 13.64



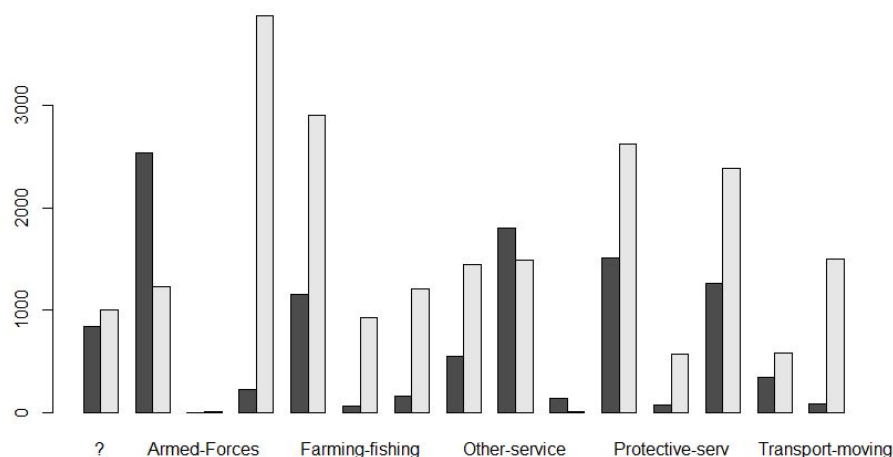
D)

We can tell a few things about these graphs most of all that we can say is that the majority of people make less than 50k, males make more income than females, and the most of the high paying jobs are in the private sector along with being some sort of upper management. There are a majority of married people but if you are single you have a higher opportunity to have more income. The age that you will get the most income is around the 30's to 40's.

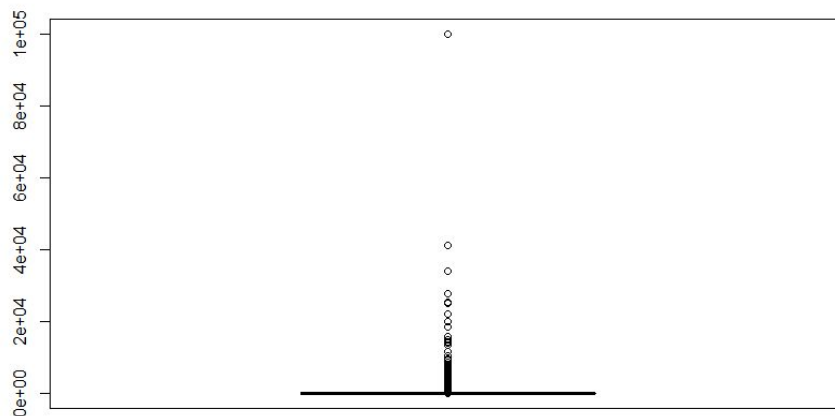
- E) The hours have an asymmetrical, non-normal, distribution because there are majority who are working around the 40 hour margin, meanwhile there are others who are working quite less than that, or quite a lot more than the 40 hours



F)



G) Yes according to the boxplot below there are a few outliers within this dataset. You can see that majority of the data points are all the way at the bottom, versus a few that are very high up and alone.



## Problem 2

a. Support =  $23/168$       Confidence =  $23/33$

There are 168 total employees, 33 of which belong to the systems department, and 23 of which are in the 46K..50K salary range.

				status	
department				junior	senior
sales	systems	marketing	secretary		
40 in 26..30	23 in 46..50	7 in 41..45	6 in 26..30	46 in 26..30	3 in 36..40
38 in 31..35	10 in 66..70	11 in 46..50	3 in 36..40	38 in 31..35	41 in 46..50
30 in 46.50	13/33 = 39%	7/18 = 38%	3/9 = 33%	7 in 41..45	10 in 66..70
b. 68/108 = 63%				23 in 46..50	13/54 = 24%
				68/114 = 60%	
age					
21..25	26..30	31..35	36..40	41..45	46..50
20 in 46..50	46 in 26..30	38 in 31..35	11 in 46..50	5 in 66..70	3 in 36..40
0/20 = 0%	3 in 46..50	7 in 41..45	0/11 = 0%	0/5 = 0%	0/3 = 0%
	3/49 = 6%	30 in 46..50			
		5 in 66..70			
		42/80 = 53%			

Note that in the above calculations, the green text indicates the most frequent (dominant) class. The calculation at the bottom of each column is the error for the given subset (complement of the dominant class/total observations in subset).

Dept:  $((68/108)*(108/168)) + ((10/33)*(33/168)) + ((7/18)*(18/168)) + ((3/9)*(9/168)) = 52.38\%$

Status:  $((13/54)*(54/168)) + ((68/114)*(114/168)) = 48.21\%$

Age:  $(0) + ((3/49)*(49/168)) + ((42/80)*(80/168)) + (0) + (0) + (0) = 26.79\%$

The last three calculations are calculating total error of the three sets (misclassification rates). The zeroes for age indicate pure sets, as shown in the trees. These percent answers make it obvious the using the input attribute age will yield the lowest misclassification rate.