

CSC 587-W1

Homework 1

Mike Turley

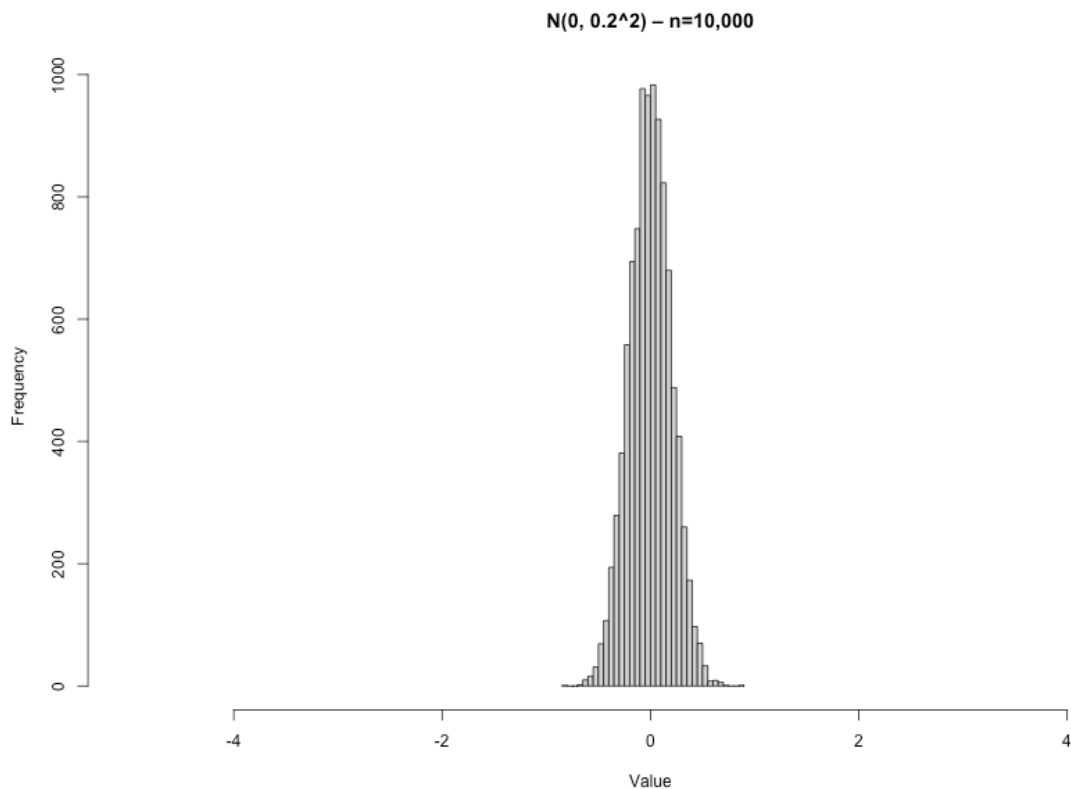
1.

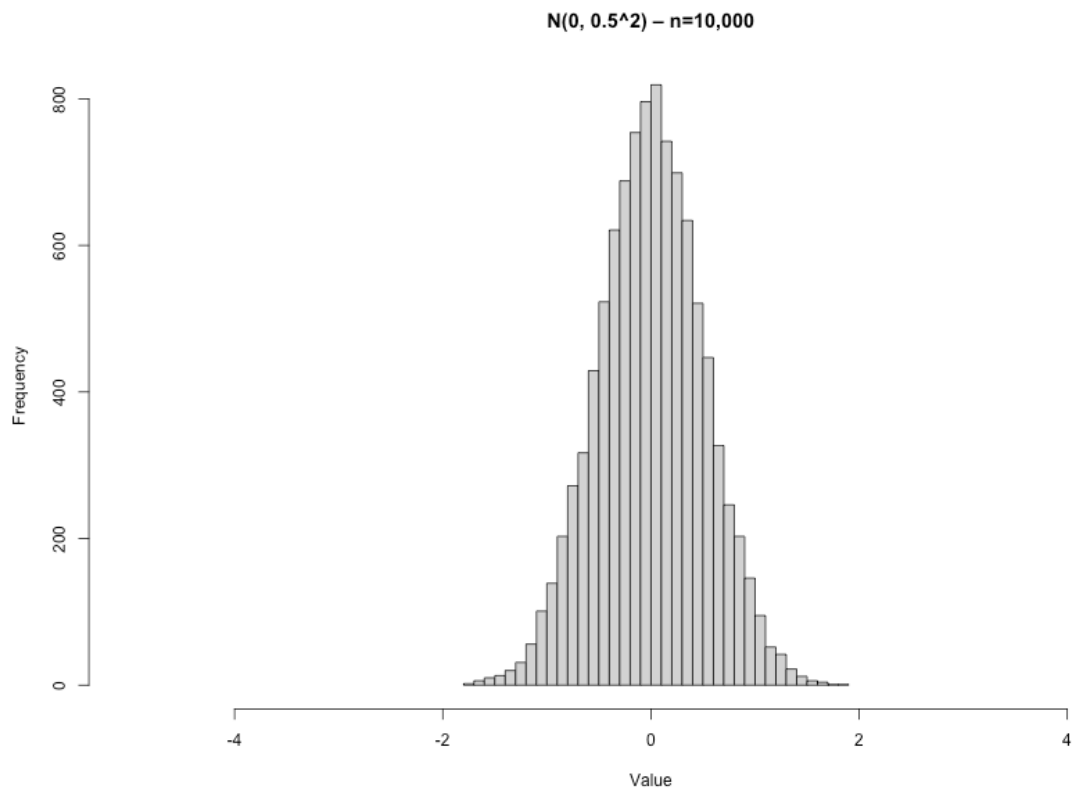
1a) Loaded Su_raw_matrix.txt into R using read.delim, confirming that the header included Liver_2.CEL.

1b) Calculated the mean and standard deviation of Liver_2.CEL to summarize its central tendency and variability.

1c) Computed column-wise means and sums for all numeric variables, giving an overview of overall expression levels across samples.

2)





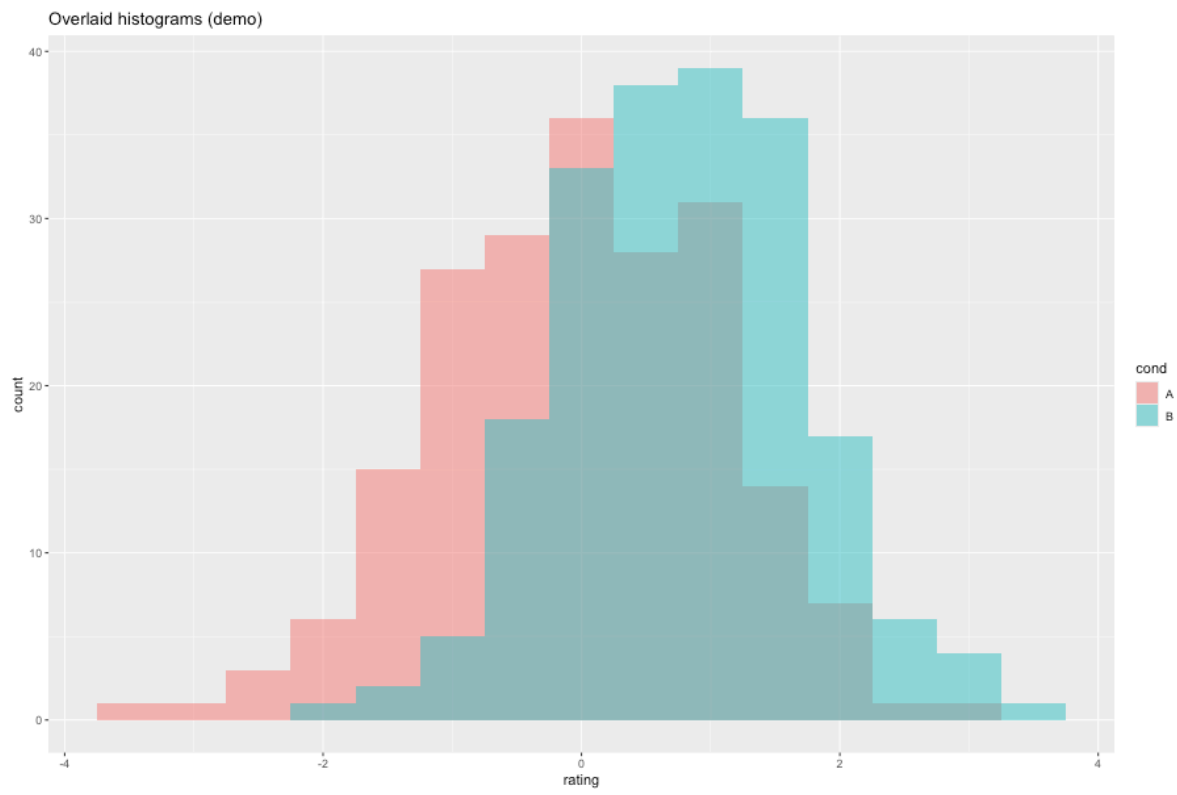
Histogram differences between `rnorm` with $\sigma = 0.2$ vs 0.5

Both sets are centered at 0 (same mean), but $\sigma = 0.5$ spreads mass more widely than $\sigma = 0.2$. With identical x-limits (i.e., -5 to 5), the $\sigma = 0.2$ histogram shows a taller, narrower shape; $\sigma = 0.5$ shows a lower, wider shape. This shows a basic property of the Normal distribution: increasing the standard deviation increases dispersion while keeping the peak at the mean.

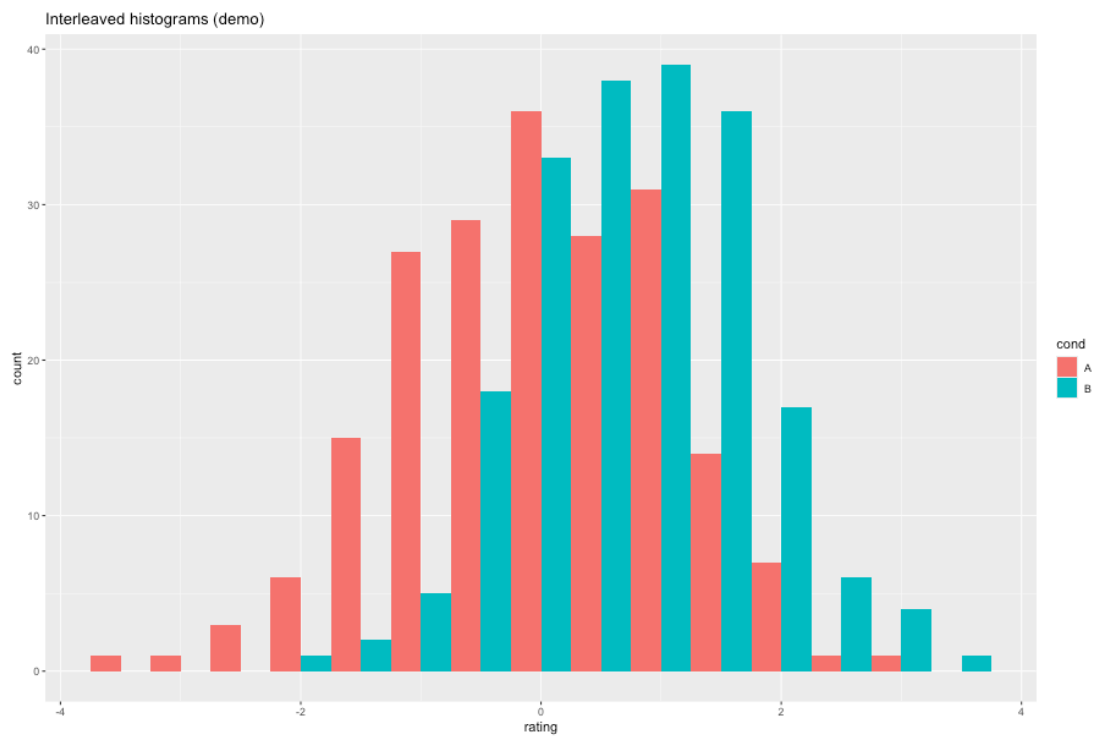
3.

3a) Create the sample dataframe

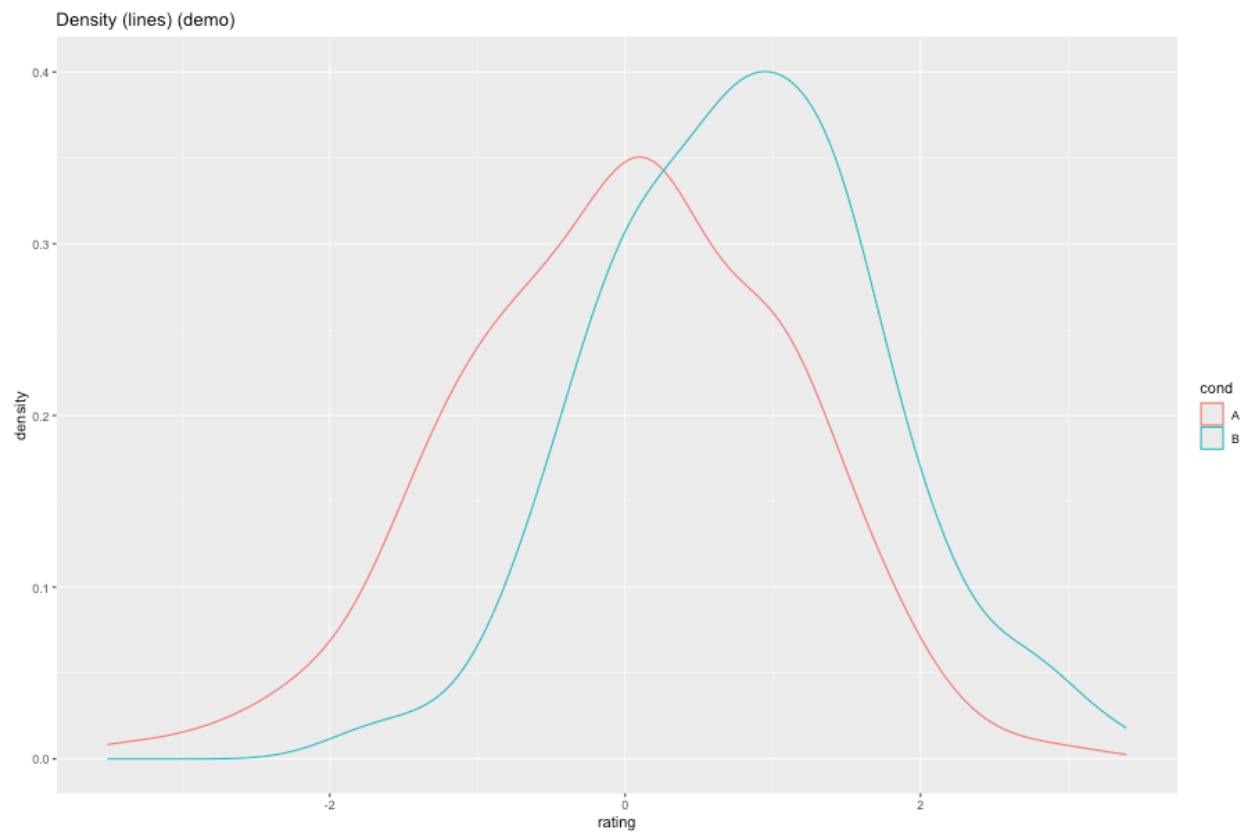
3b)



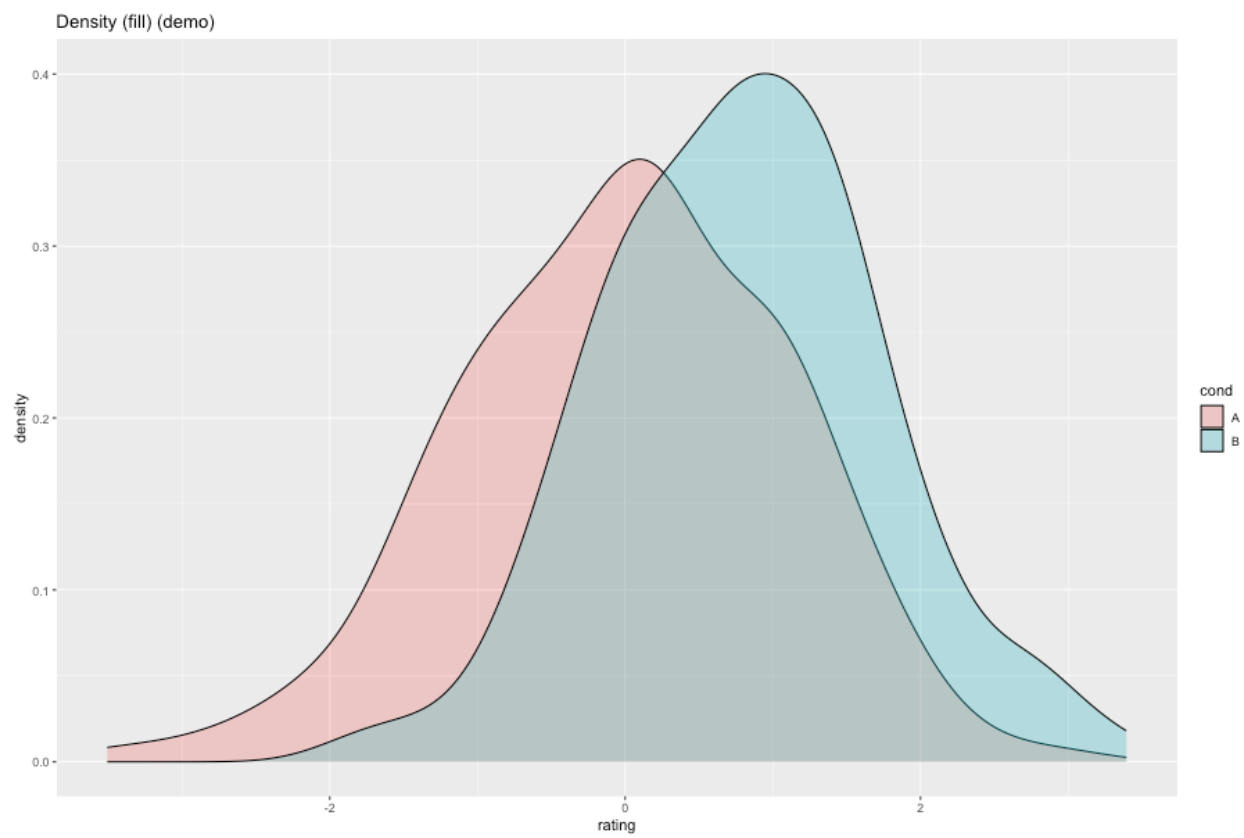
3c)



3d)



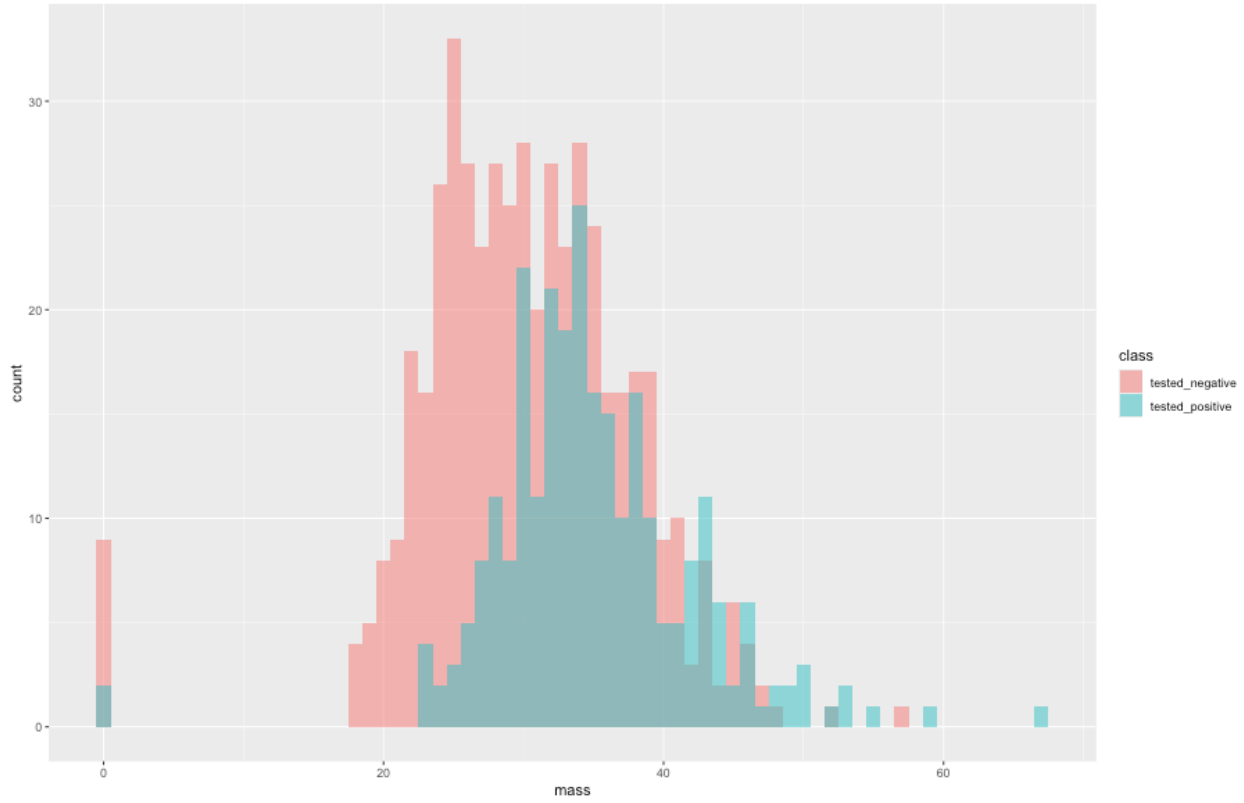
3e)



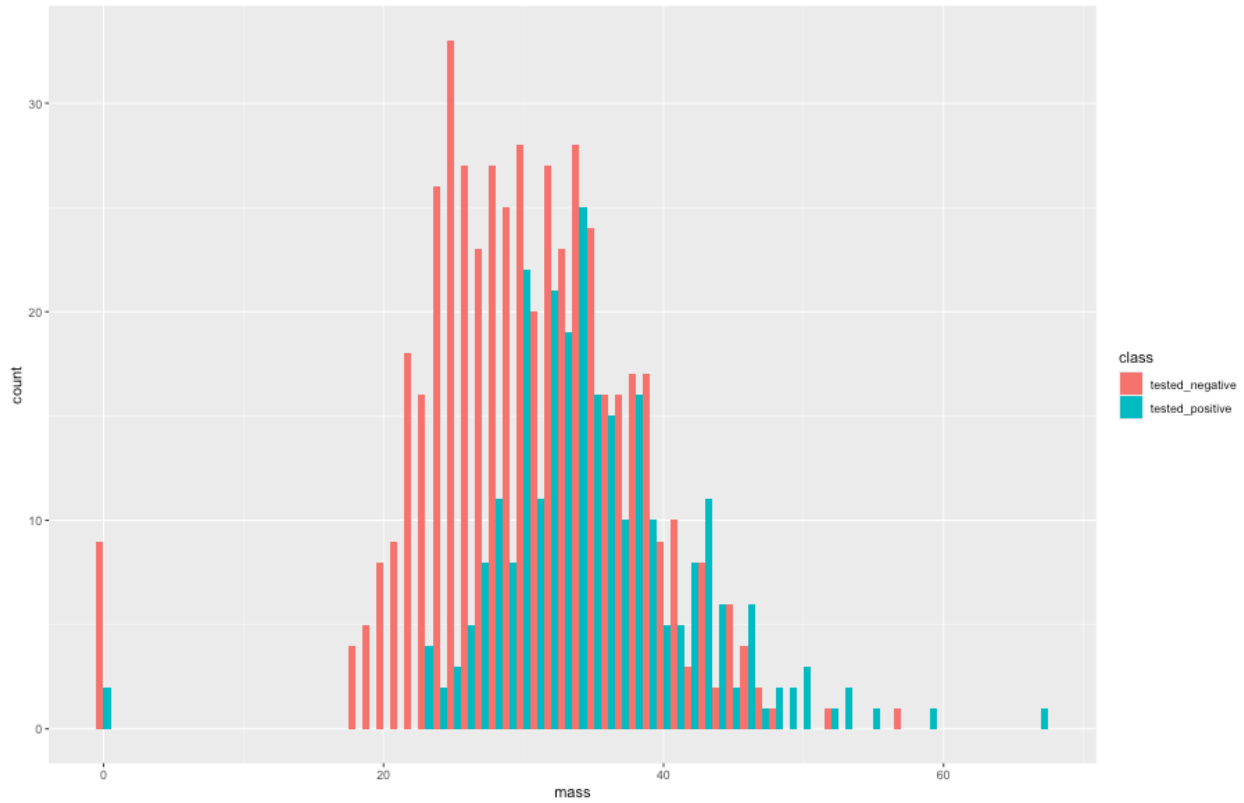
3f) Diabetes mass by class

Interleaved separates bars by class; overlaid stacks them in the same x-bins (transparency helps).

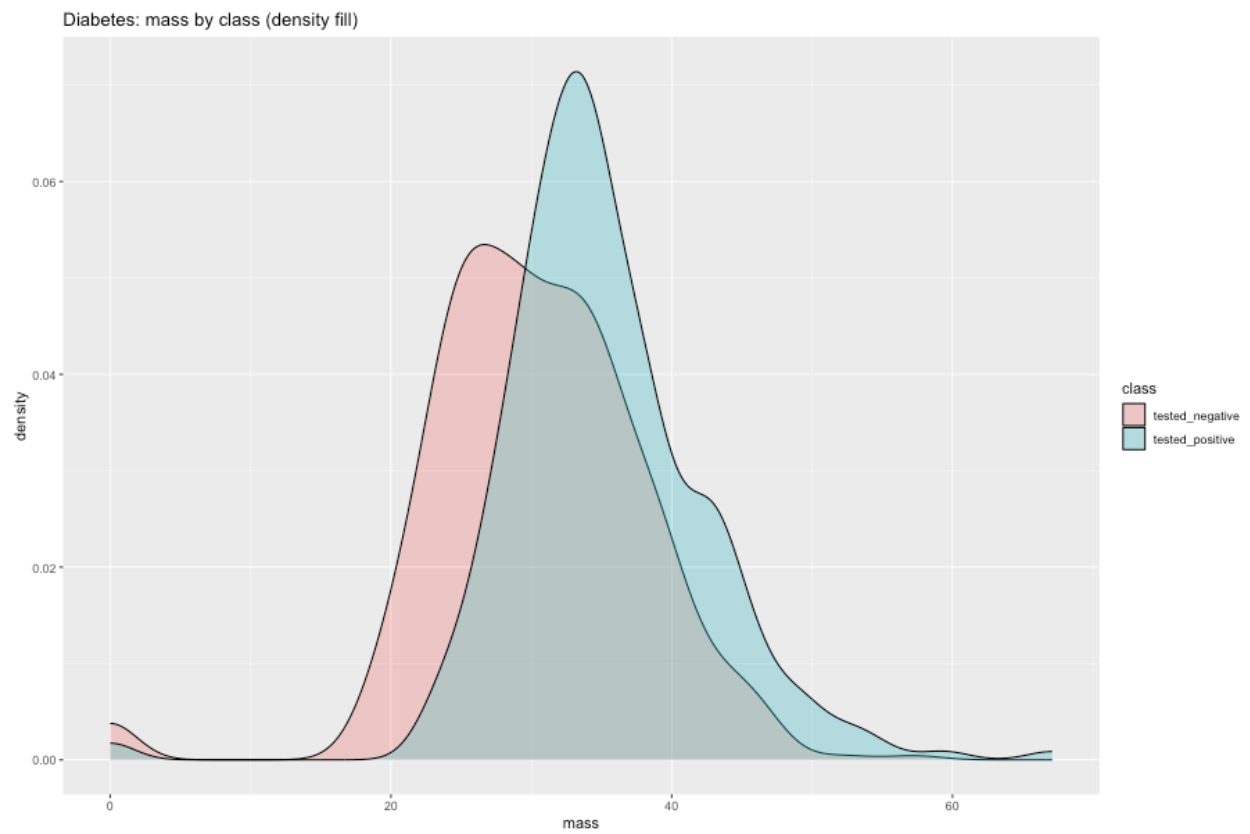
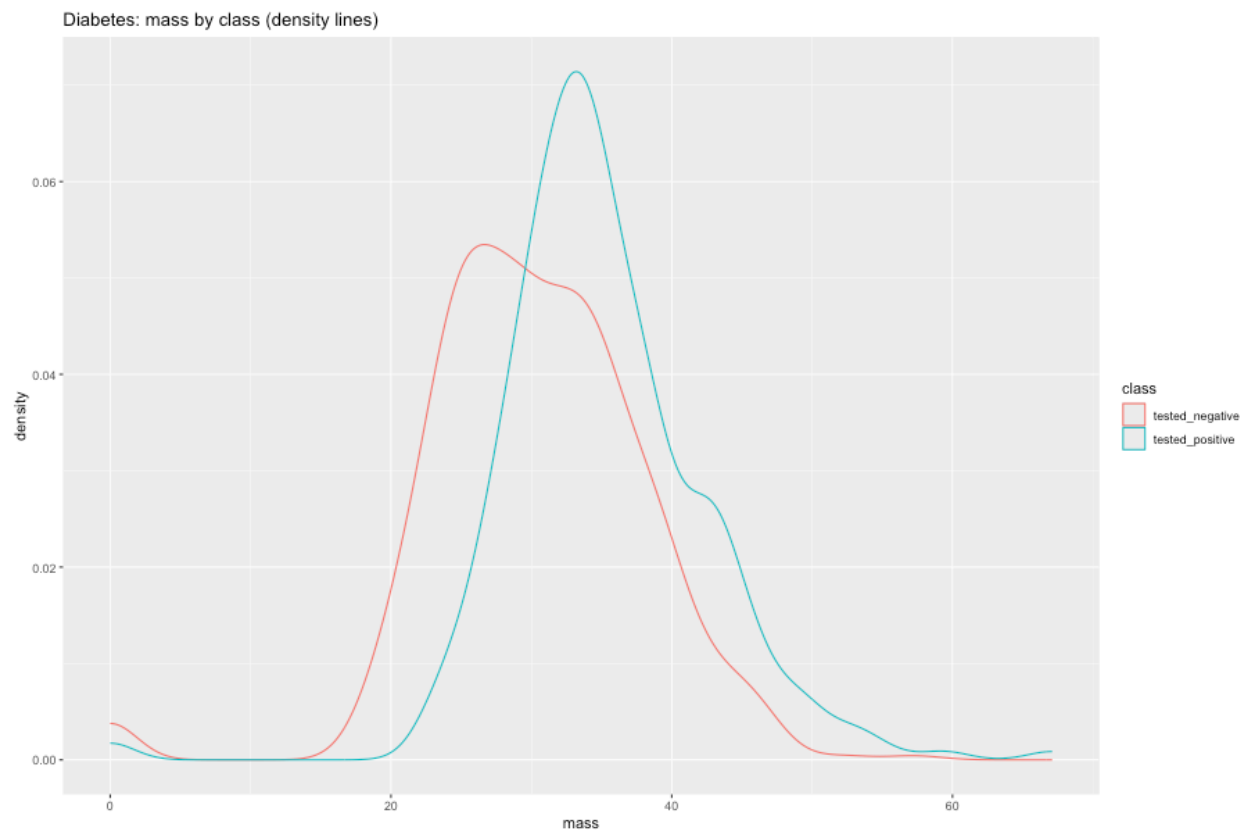
Diabetes: mass by class (overlaid)



Diabetes: mass by class (interleaved)



For the density the lines emphasize distribution shape; filled density with alpha visually conveys overlap and separation between classes.



4.

4a) `drop_na() %>% summary()`: removes rows with missing values, then prints univariate summaries, ensuring stats aren't biased by NAs.

4b) `filter(Sex == "male")`: subsets to male passengers.

4c) `arrange(desc(Fare))`: sorts passengers by descending fare (highest first).

4d) `mutate(FamSize = Parch + SibSp)`: creates Family Size as sum of parents/children and siblings/spouses.

4e) `group_by(Sex) %>% summarise(...)`: aggregates by sex to report mean fare and number of survivors.

5.

Used the `quantile()` function in R to compute the **10th, 30th, 50th, and 60th percentiles** of the `skin` variable in the diabetes dataset. These values highlight key points in the distribution and help us understand how skin thickness is spread across patients.

