

8.1) Assuming normal distribution, mean $\mu = 242$ months, standard deviation (σ) = 8 months.

We have to find $P(20 < x < 30)$.

$$\text{Normal distribution, } f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

Normalizing the variables,

$$z = \frac{(x - \mu)}{\sigma} \quad \therefore z_1 = (20 - 242) / 8 = -2.75$$

$$z_2 = (30 - 242) / 8 = -12 / 8 = -1.5$$

$$\begin{aligned} \therefore P(20 < x < 30) &= P(-2.75 < z < -1.5) \\ &= P(z < -1.5) - P(z < -2.75) \end{aligned}$$

from normal distribution table,

$$P(z < -1.5) = 0.0668.$$

$$P(z < -2.75) = 0.0030.$$

$$\begin{aligned} \therefore P(20 < x < 30) &= 0.0668 - 0.0030 \\ &= 0.0638 \end{aligned}$$

Q.2) Since, we have a 100 + data, so the ~~med~~ mean cannot be calculated, but as ~~the~~ here median will be a useful measure, as data is readily available. Here, there are 8 data points, therefore ~~mean~~ median = Mean of 4th and 5th data points = $(63 + 75) / 2 = 69$.

Again here 8 different parts of an electronic component is given, so, the lowest value 36 will also be a useful measure as failure of a single part will ~~may lead to~~ ~~result~~ ~~for~~ failure of whole electronic component.

Q.3) For
= 10.75
2.53,
500
sample

sample

= 10

∴ Sam

For

$Q_1 =$

$Q_2 =$

Q.3) First formulation time data points
= 1.75, 1.92, 2.62, 2.35, 3.09, 3.15,
2.53, 1.91

See Here $n = 8$.

$$\text{Sample Mean} = \frac{\sum \text{times}}{n} = \mu$$

$$= \frac{(1.75 + 1.92 + 2.62 + 2.35 + 3.09 + 3.15 + 2.53 + 1.91)}{8}$$

$$= 19.32 / 8 = 2.415$$

$$\text{Sample Variance} = \frac{\sum (x_i - \bar{x})^2}{(n-1)} = s^2$$

$$= 10. \quad \approx (1.997617)$$

$$\approx 0.285371$$

$$\therefore \text{Sample standard deviation} = \sqrt{\text{Var}(n)} = s$$

$$= \sqrt{0.285371} = 0.5342$$

For the first data set,

$$\min = 1.75$$

$$Q_1 = (n+1)/4 = 2.25 \rightarrow \text{between 2nd and 3rd} \\ = (1.91 + 1.92)/2 = 1.915$$

$$Q_2 = \text{median} = \text{between 4th and 5th} \\ = (2.35 + 2.53)/2 = 2.44$$

$$\text{Similarly } Q_3 = 2.9725$$

$$\text{Max} = 3.15$$

Similarly for 2nd dataset.

$$\text{Min} = 1.83$$

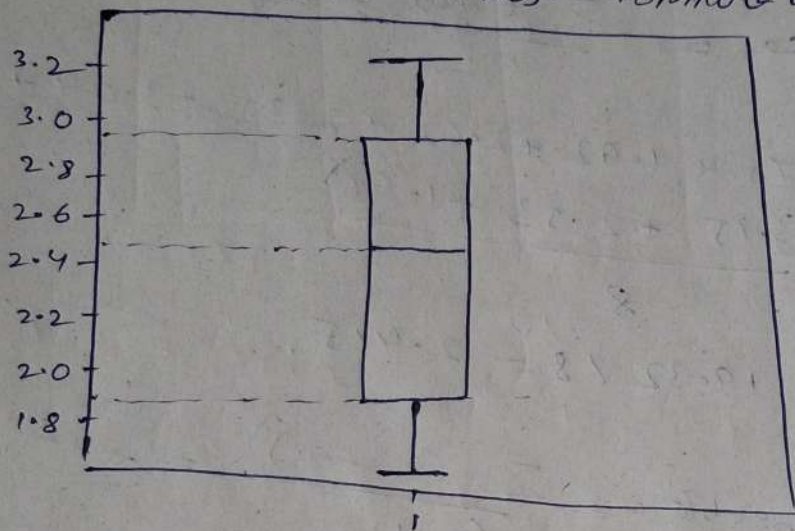
$$Q_3 = 3.305$$

$$Q_1 = 1.965$$

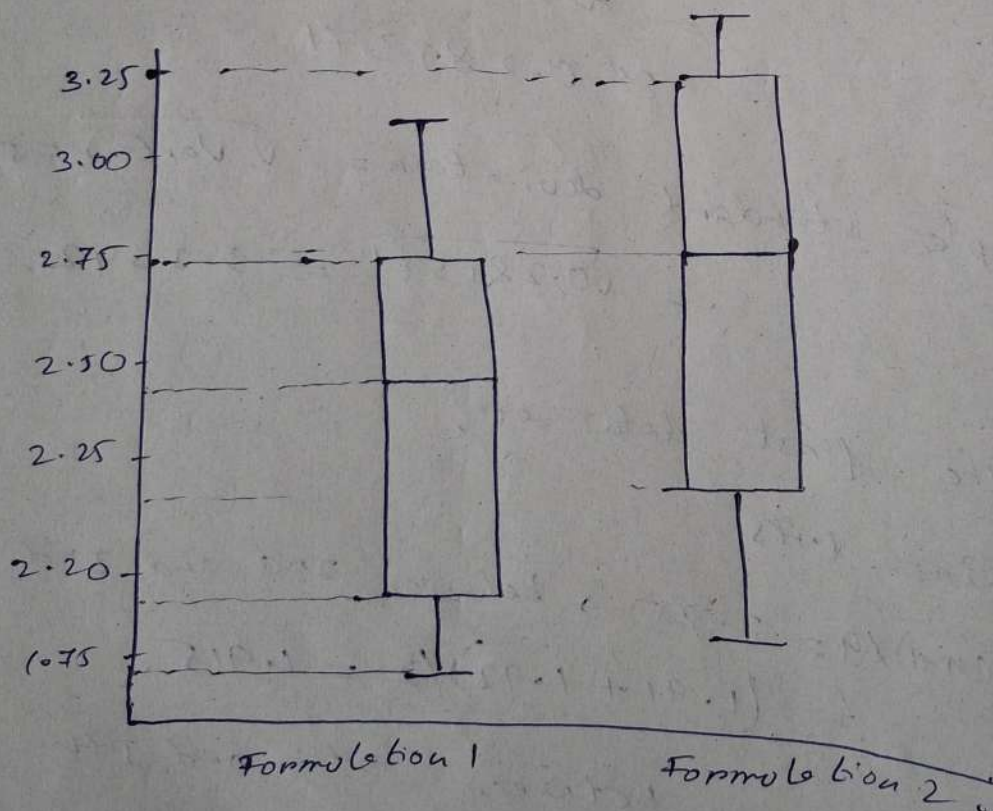
$$\text{Max} = 3.40$$

$$\text{Median} = 2.76$$

cold start Times - Formulation - 1.



Comparative cold start Times.



→ Neither box plot points plotted beyond the whiskers, indicating no outliers by 1.5 x IQR rule.

Conclusion

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with a
Formulation
and gr.
If quick
I appear

Conclusion :- Formulation 1 offers faster and more consistent cold start ignition times, with a lower median and tighter spread. Formulation 2, exhibits slower average and greater variability.

If quick, reliable is priority, - Formulation 1 appears superior.

4) For first ~~two~~ entries let's analyze

a) Min-max weight normalization

$$\text{formule} = \frac{\text{weight}(w_i) - \text{Min}(\text{min}_i)}{\text{Max}(\text{max}) - \text{Min}}$$

$$= \frac{50 - 41}{136 - 41} = 0.094737$$

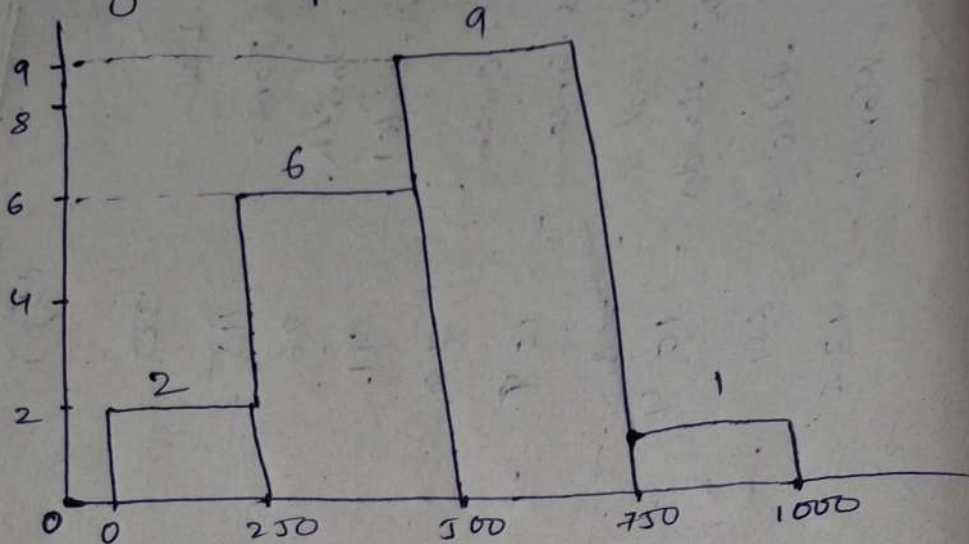
b) P. lee has weight 50kg which is less than 60kg. Therefore his/her weight falls in low category.

$$\begin{aligned} \text{c) BMI} &= \frac{\text{weight (kg)}}{(\text{Height (m)})^2} = \frac{50 \text{ kg}}{(1.52)^2 \text{ m}^2} \\ &= \cancel{2.16} \quad 21.641274 \text{ kg/m}^2 \end{aligned}$$

Table with all the newly added columns.

Name	Weight (kg)	Height (m)	BMI (kg/m ²)	Normalized Weight	Weight Category	Blood Pressure	Diabetes
1. P. Lee	50	1.052	21.641274	0.094737	Low	68/112	0
2. R. Jones	115	1.0177	36.707204	0.778947	High	110/152	0
3. J. Smith	96	1.083	28.666129	0.578947	Medium	88/132	0
4. A. Patel	41	1.0155	17.065557	0.0	Low	76/125	0
5. M. Owen	79	1.082	23.849777	0.40	Medium	65/105	0
6. S. Green	109	1.089	30.514263	0.715789	High	101/132	0
7. N. Cook	73	1.071	23.566632	0.336842	Medium	108/136	0
8. W. Harris	104	1.074	35.566499	0.663158	High	101/132	0
9. P. Rice	64	1.074	21.0138856	0.242105	Medium	101/132	0
10. F. Marsh	136	1.078	42.923873	1.00	High	121/165	1

5. a) Histogram of sale Price.



b) Contingency Table: Store vs. Product Category

<u>Store</u>	<u>Desktop</u>	<u>Laptop</u>	<u>Printer</u>	<u>Scanner</u>
New York, NY	3	1	2	4
Washington, DC	2	2	2	2

c) Summary Table - Grouped by Customers.

<u>Customer</u>	<u>Transaction Count</u>	<u>Total Sale Price (\$)</u>
B. March	3	1700
E. Sims	1	700
G. Hinton	4	2150
H. Fu	1	450
H. Taylor	1	400
J. Baird	1	500
L. Nye	2	900
P. Todd	2	900
S. Cann	1	600
Total	2	750

ii) Summary

Store
New York
Washington

iii) Grouped

Product

Desktop
Laptop
Printer

d) Scanner

Profit (\$)

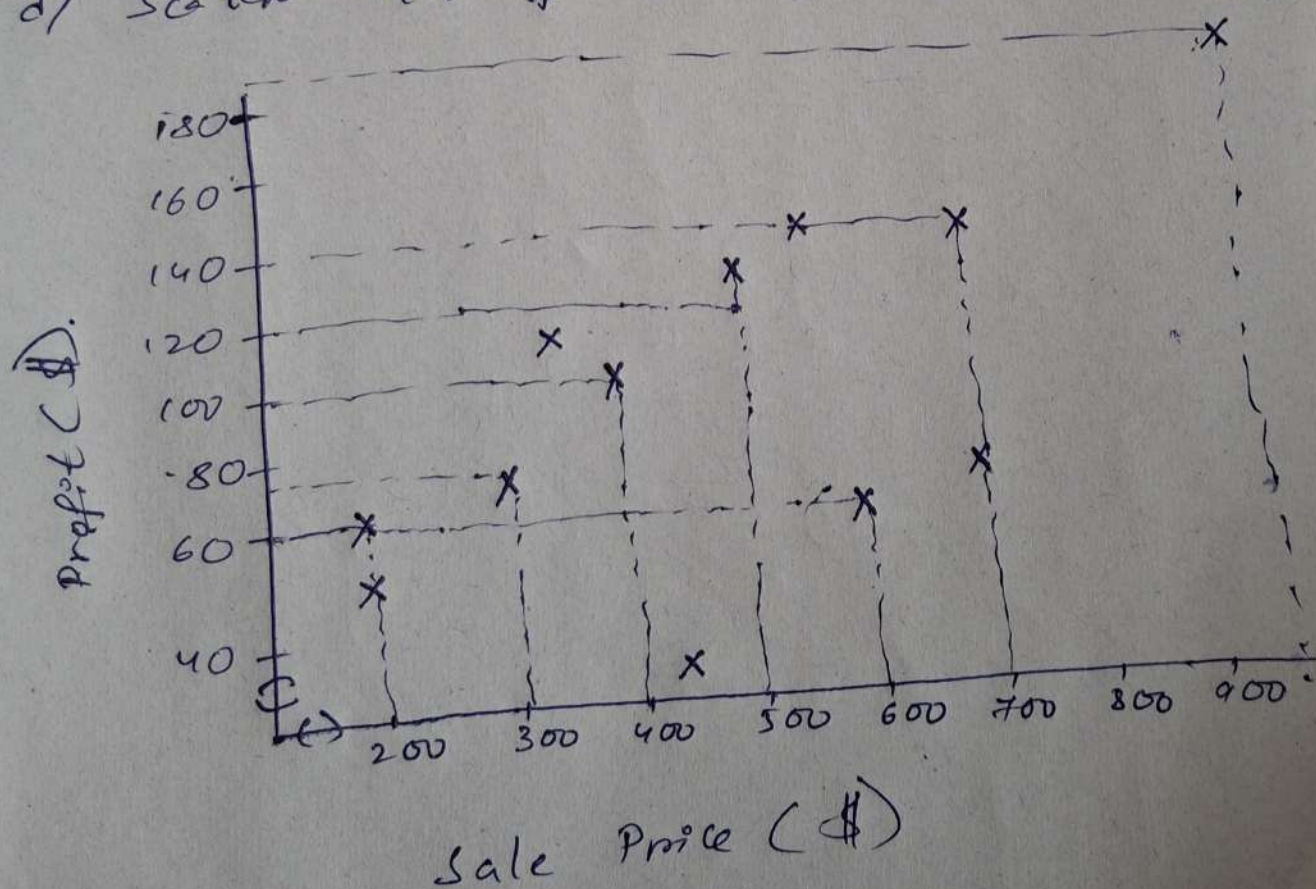
ii) Summary Table ii - Grouped by store.

<u>store</u>	<u>Transaction Count</u>	<u>Mean Sale Price (\$)</u>
New York, NY.	10	485.0
Washington DC.	8	525.0

iii) Grouped by Product Category.

<u>Product Category</u>	<u>Transaction Count</u>	<u>Total Profit (\$)</u>
	5	295
Desktop.	3	470
Laptop.		360
Printer.	4	640
Scanner.	6	

d) Scatter Plot of Sale Price vs Profit.



Frequency of samples for each class:

Class

A 151

B 123

C 68

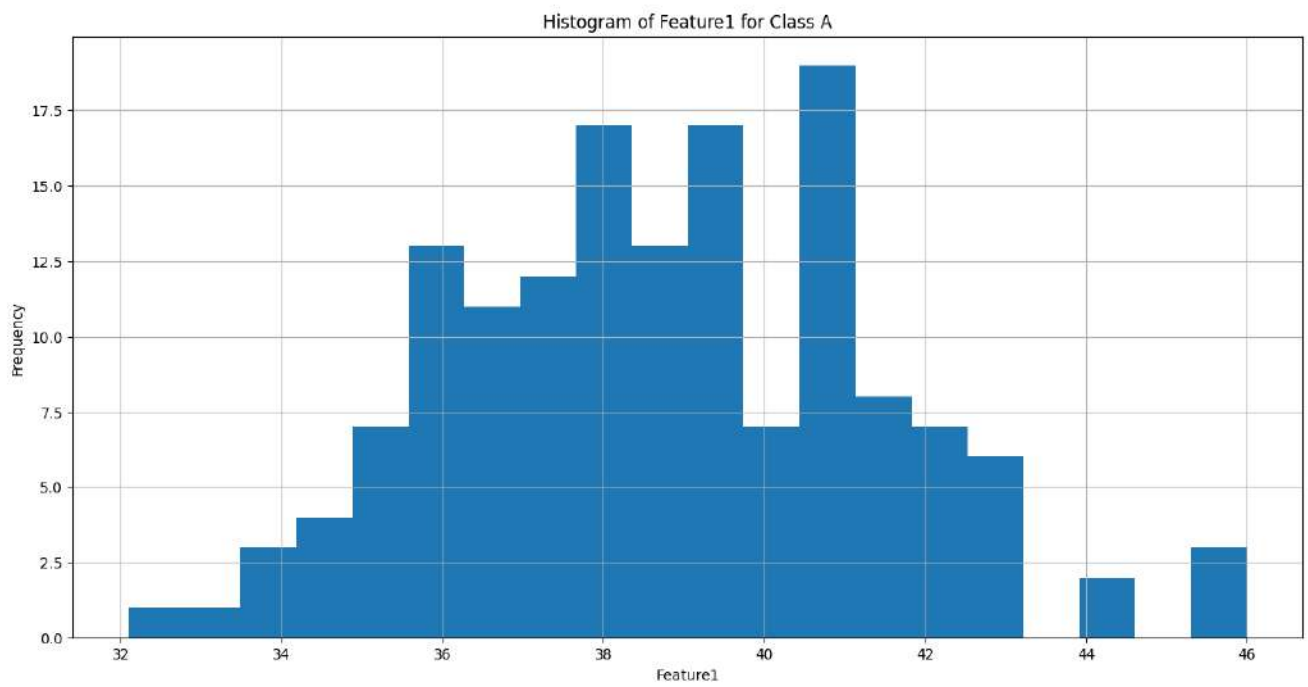
Name: count, dtype: int64

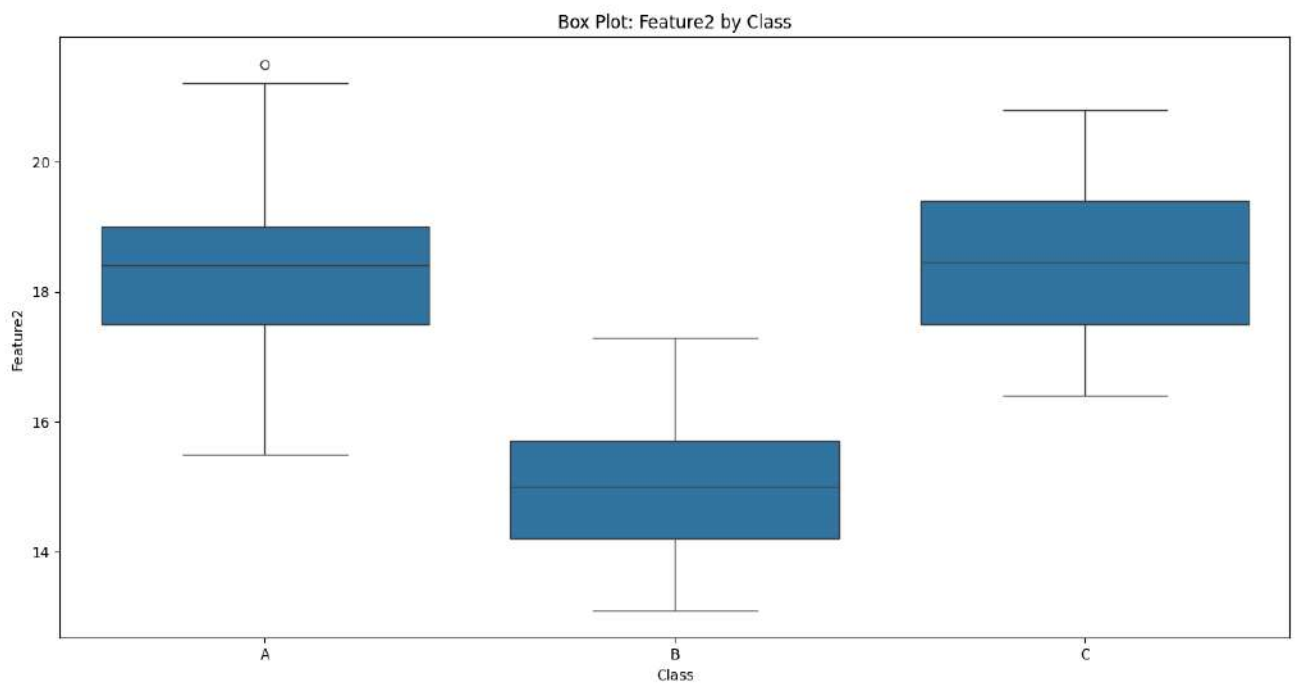
Data Description:

	Feature1	Feature2	Feature3	Feature4
count	342.000000	342.000000	342.000000	342.000000
mean	43.921930	17.151170	200.915205	4201.754386
std	5.459584	1.974793	14.061714	801.954536
min	32.100000	13.100000	172.000000	2700.000000
25%	39.225000	15.600000	190.000000	3550.000000
50%	44.450000	17.300000	197.000000	4050.000000
75%	48.500000	18.700000	213.000000	4750.000000
max	59.600000	21.500000	231.000000	6300.000000

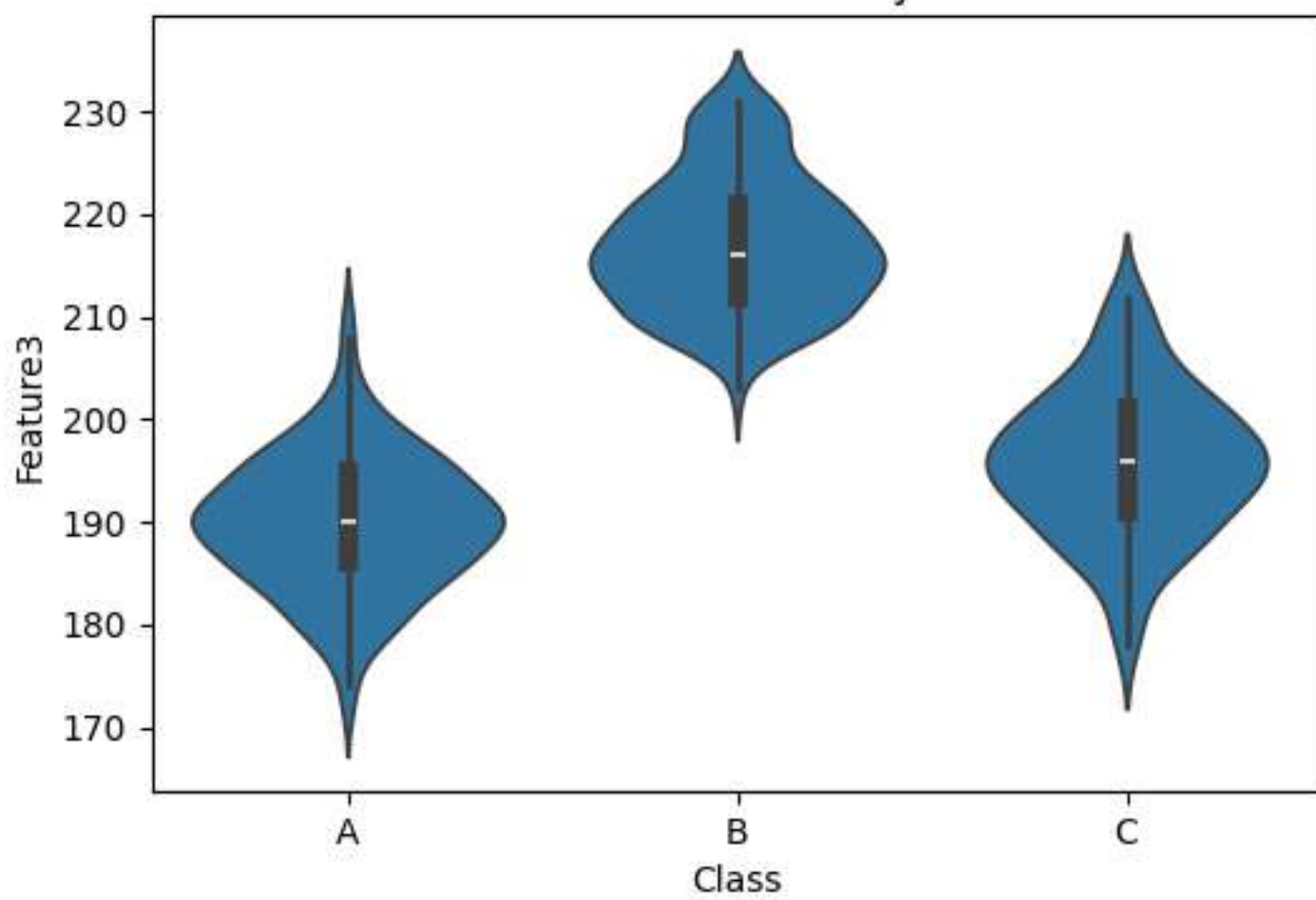
Interquartile Range (IQR):

Feature1	9.275
Feature2	3.100
Feature3	23.000
Feature4	1200.000
dtype: float64	

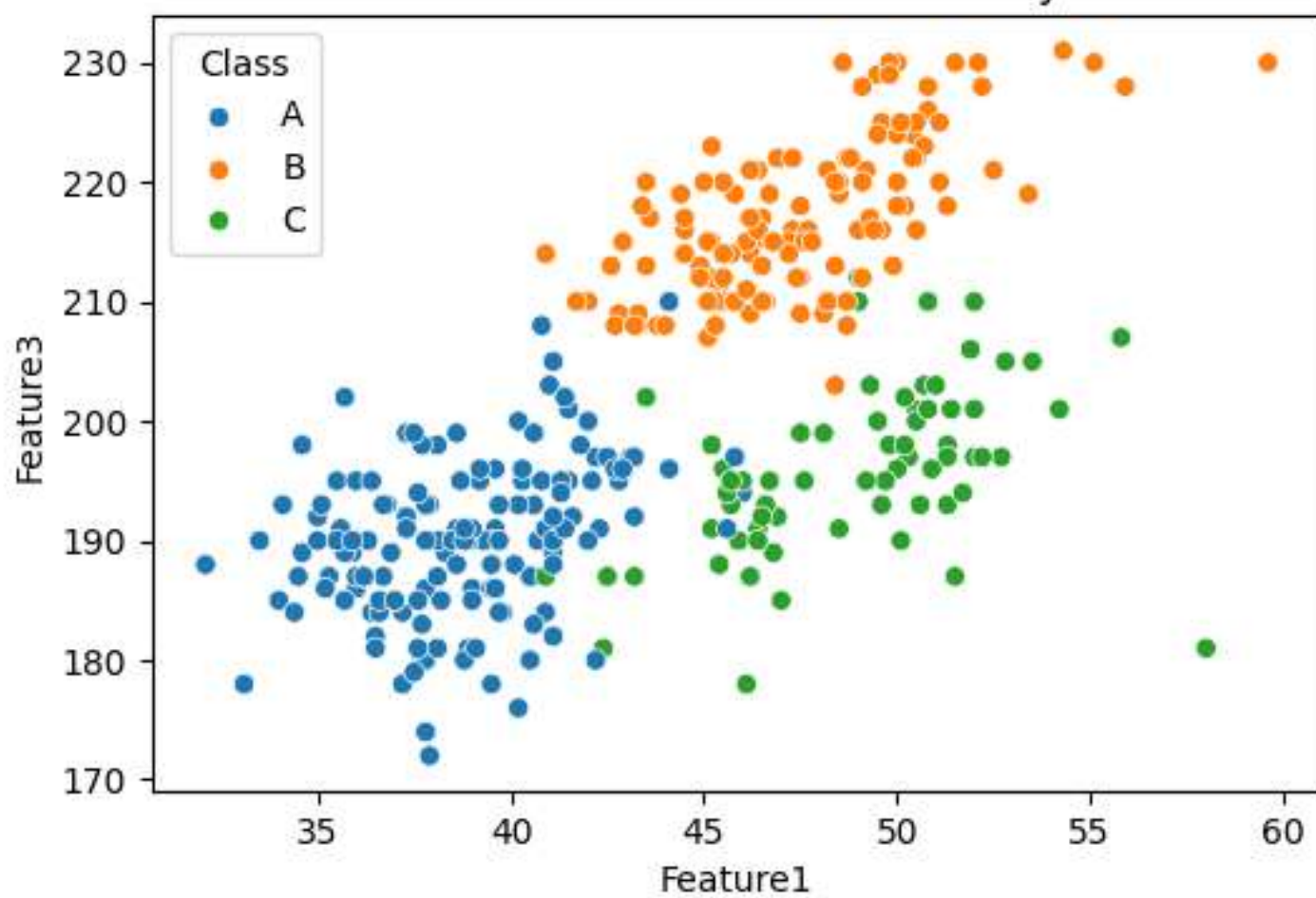




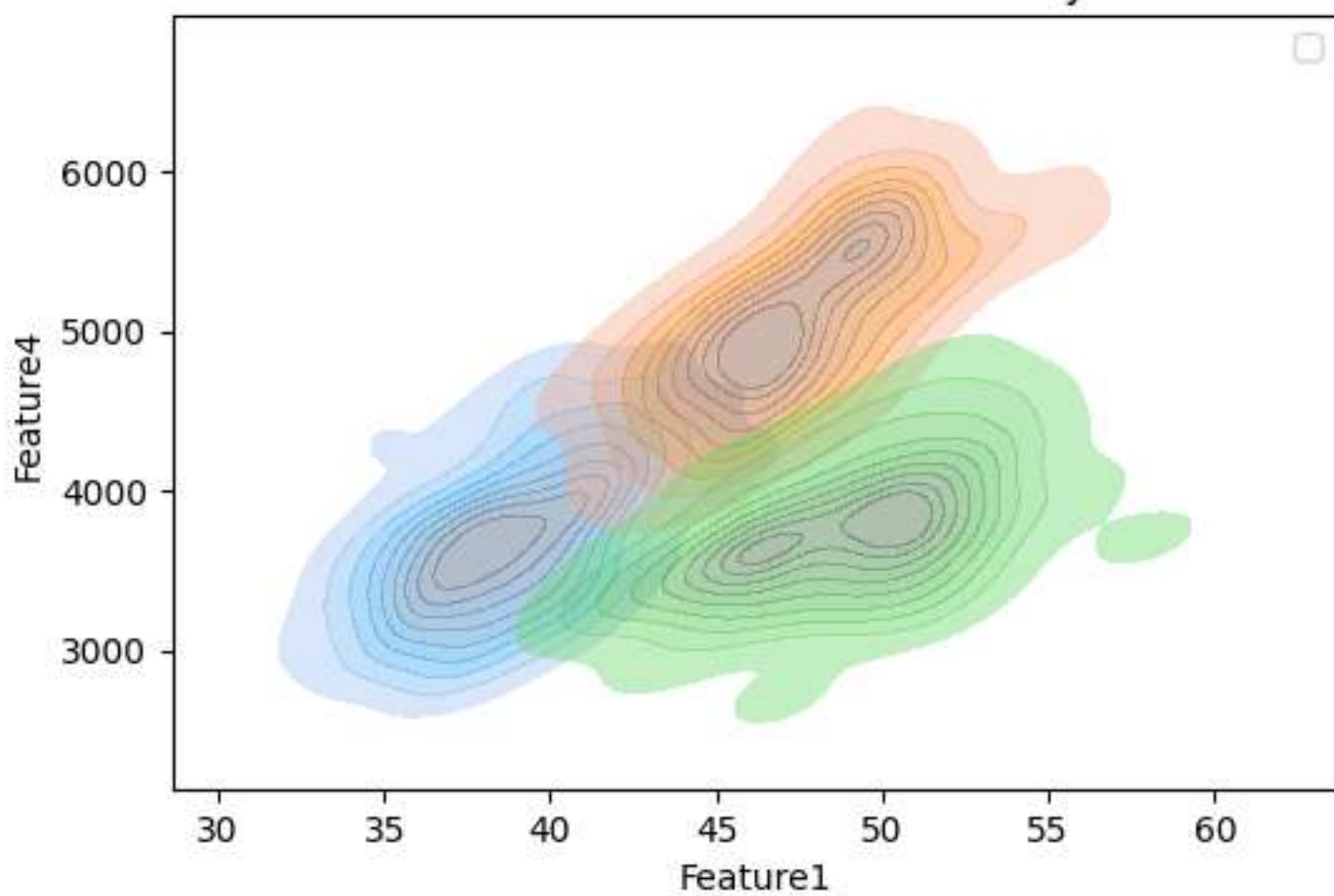
Violin Plot: Feature3 by Class



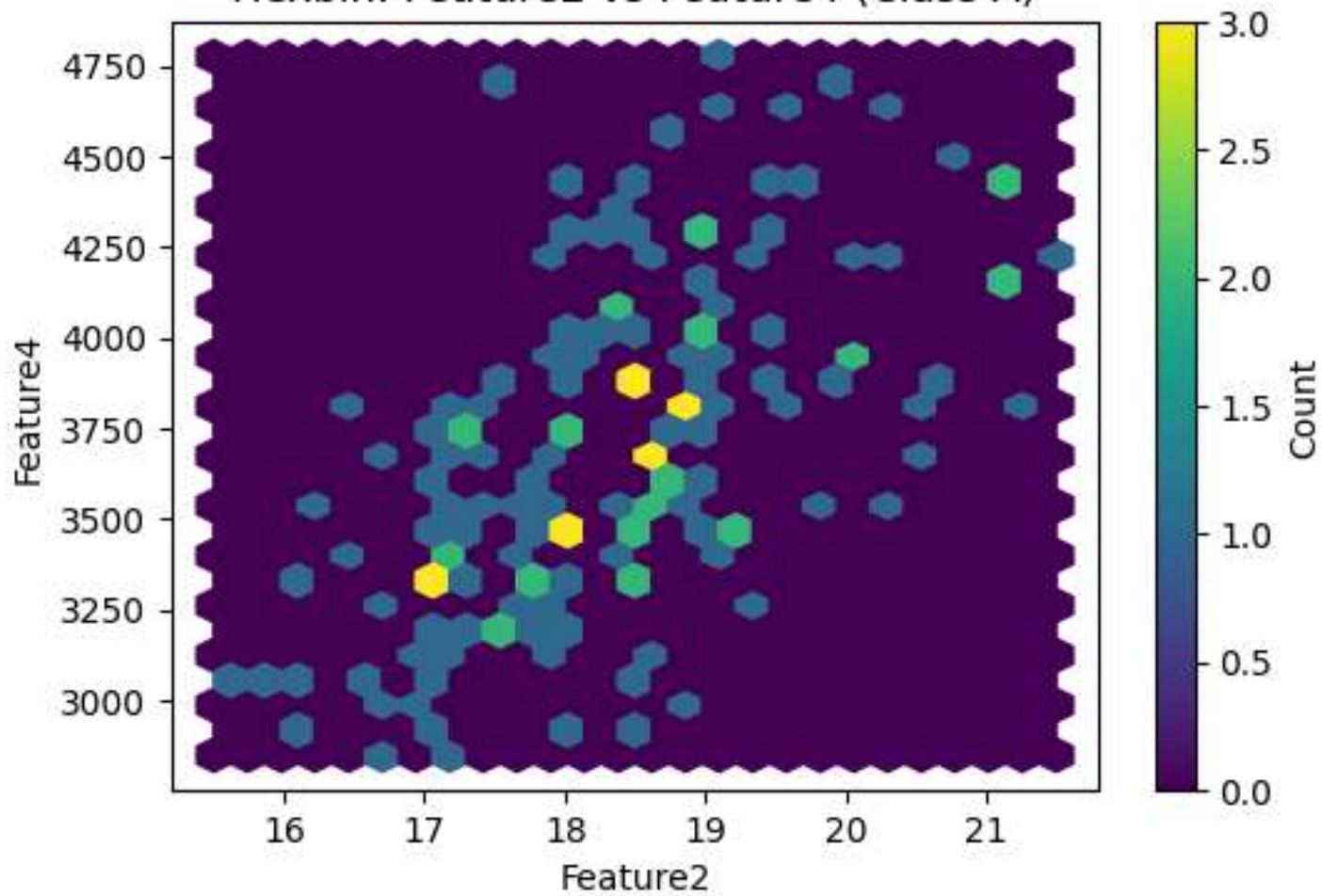
Scatter Plot: Feature1 vs Feature3 by Class



Contour Plot: Feature1 vs Feature4 by Class



Hexbin: Feature2 vs Feature4 (Class A)



Correlation Matrix

