1.

- a. The K-means clustering algorithm is sensitive to outliers, because a mean is easily influenced by extreme values. ... Mean is greatly influenced by the outlier and thus cannot represent the correct cluster center, while medoid is robust to the outlier and correctly represents the cluster center
- b. Will define: $J = \sum_{i=1}^{m} (x_i \mu)^2$

Will differentiate with respect to μ

And get:
$$\frac{\partial J}{\partial \mu} = \sum_{i=1}^{m} -2(x_i - \mu)$$

Will substitute
$$\mu = \frac{\sum_i^m x_i}{m}$$
 and will get $= \sum_{i=1}^m -2\left(x_i - \frac{\sum_i^m x_i}{m}\right) = \left(-2(mx_1 - x_1 - x_2 - \cdots x_m) + \cdots + -2(mx_m - x_1 - x_2 - \cdots x_m)\right)$ will devide by -2 and than will get $= (m(x_1 + \cdots + x_m) - mx_1 - \cdots - mx_m) = 0$

Therefor it's a extremum point will show it's a minimum rather than a maximum

Will differentiate again: $\frac{\partial^2 J}{\partial \mu^2} = 2m\mu > 0$ so the 2nd derivative is always positive and thus the function is convex and have only minimum point.

c. Will define $J_2 = \sum_{i=1}^m |(x_i - \mu)|$

$$f(x) = |x| \rightarrow f'(x) = \frac{x}{|x|}$$
 therefor

Will differentiate with respect to μ

And get:
$$\frac{\partial J}{\partial \mu} = -\sum_{i=1}^m \frac{(x_i - \mu)}{|(x_i - \mu)|}$$

Will substitute
$$\mu = median(x)$$
 and will get $\frac{\partial J}{\partial \mu} = -\sum_{i=1}^{m} \frac{(x_i - median(x))}{|(x_i - median(x))|} =$

Will open:
$$-\frac{(x_1 - median(x))}{|(x_1 - median(x))|} - \dots - \frac{(x_m - median(x))}{|(x_m - median(x))|} =$$
$$-1 \left(\frac{\left(x_1 - median(x)\right)}{|(x_1 - median(x))|} + \dots + \frac{\left(x_m - median(x)\right)}{|(x_m - median(x))|} \right)$$

We know that exactly half of the x's bigger than the median and exactly half of them smaller therefore will get exactly half of the numerator are +1 and exactly half are -1 And all of the denominator are +1 there for will get something like that:

$$-1(1+1+\dots+(-1)+(-1))=-1(0)=0$$

Will differentiate again: $\frac{\partial^2 J}{\partial \mu^2} = 0$ so the 2nd derivative is always positive(or 0) and thus the function is convex and have only minimum point.

2. We have 2 SVM with linear karnel A and D . A as samples within the margins and thus have more soft margins and thus smaller penalty term C.

A-1

D-2

C and F are polynomial karnel, and its clear F is with higher degree.

F-4

C-3

B and E are both rbf, and B boundary is tighter so its with bigger γ , which is correlated with tighter boundary.

B-6

E-5

3.

- a. The scientific term is generalization
- b. 2p could imply about overfitting (lack of generalization) model if its 2 big.
 And 2ln(L) could imply about underfitting (lack of generalization) if its to small.
 Therefor a good AIC is if there is a balance between the 2 parameters.
- c. If the balance is violated there could be under/overfitting.
- d. AIC is a criterion that help you asses goodness of a model. Lower values of the index indicate the preferred model, that is, the one with the fewest parameters that still provides an adequate fit to the data.