Hand in on Gradescope before 22:00 on Jan. 30 (Saturday). Each question will be given 1, 0.5 or 0 points as follows. If the question is more or less correct it gets 1 point. If it is partly correct it gets 0.5, and if it is missing or completely wrong it gets 0 points.

1)

An algorithm for *fault detection* takes an unlabeled data set as input, $\{x^{(1)}, x^{(2)}, \dots x^{(n)}\}$ and provides a decision rule for classifying examples x as faulty or not. A simple fault detection algorithm solves the following optimization problem

$$\min_{R,c} R^2$$

subject to
$$||x^{(i)} - c||^2 \le R^2$$
, $i = 1, ... n$.

This corresponds to finding the radius R of the smallest sphere with center c that contains all the points in the training set. A point x is classified as not faulty if it lies inside the sphere, and faulty otherwise.

The above algorithm is sensitive to outliers in the sense that a single point can increase the radius significantly. Modify the above algorithm so that it becomes more resistant to outliers. *Hint*: Recall soft vs hard-margin SVM.

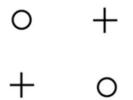
I would use the SVM with a Radial Basis Kernel.

$$min_{R,c,\gamma}R^2$$

subject to
$$e^{-\gamma ||x^{(i)}\text{-}c||^2} \ <= \ R^2$$
 , $i \ = \ 1, \dots, n.$

where γ controls the decision boundary.

Consider the following data set



Circle all the classifiers that can achieve zero training error on this data set.

- i) Logistic regression
- ii) SVM with a RBF kernel.
- iii) 3-NN classifier.

only SVM with a RBF kernel.