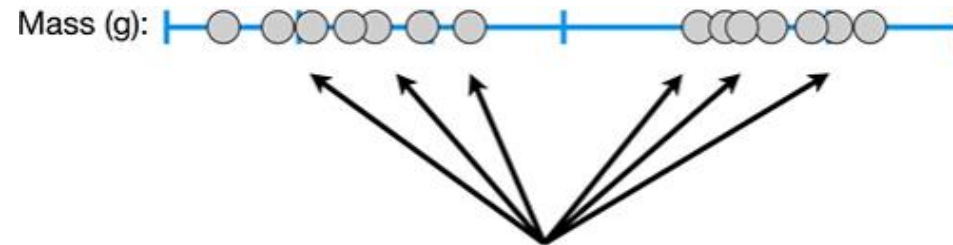


Introduction to Support Vector Machines

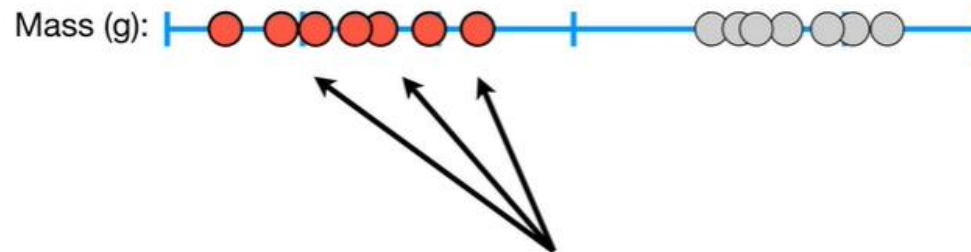
Support Vector Machine (SVM)

- SVM is a supervised [machine learning algorithm](#) that can be used for both classification or regression problem.
- In the SVM algorithm, we plot each data item as a point in n-dimensional space (where n is a number of features you have) with the value of each feature being the value of a particular coordinate.
- Then, we perform classification by finding the hyper-plane that differentiates the two classes very well.

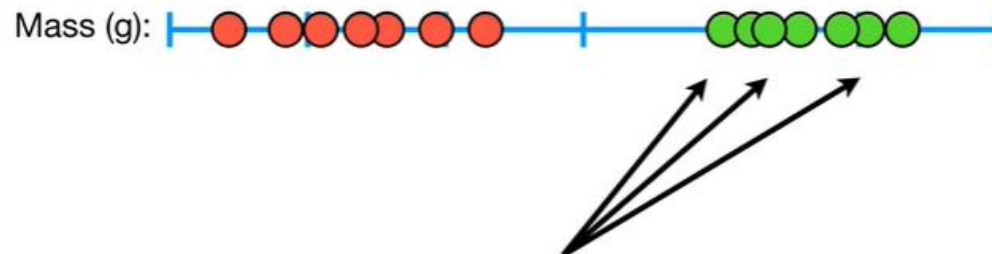
SVM



Let's start by imagining we measured
the mass of a bunch of mice...

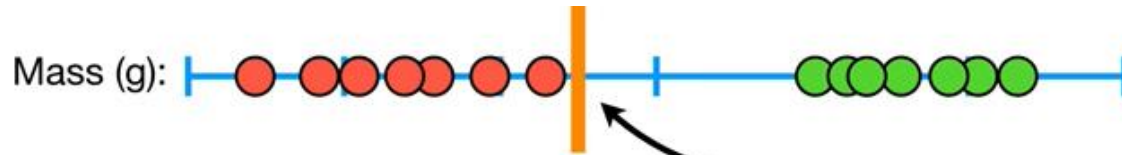


The **red dots** represent mice are **not obese**...

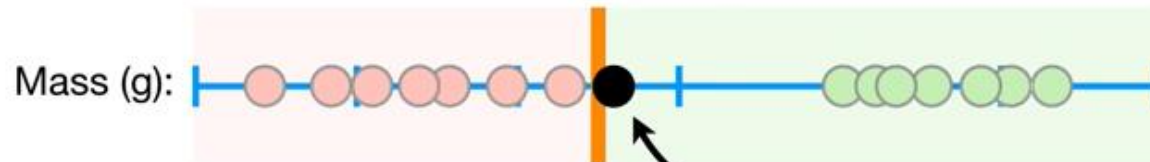


...and the **green dots** represent mice are **obese**.

SVM



Based on these observations, we can pick a threshold...



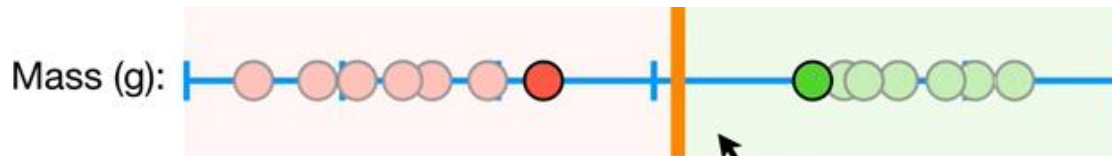
However, what if get a new observation here?

SVM

Mass (g):

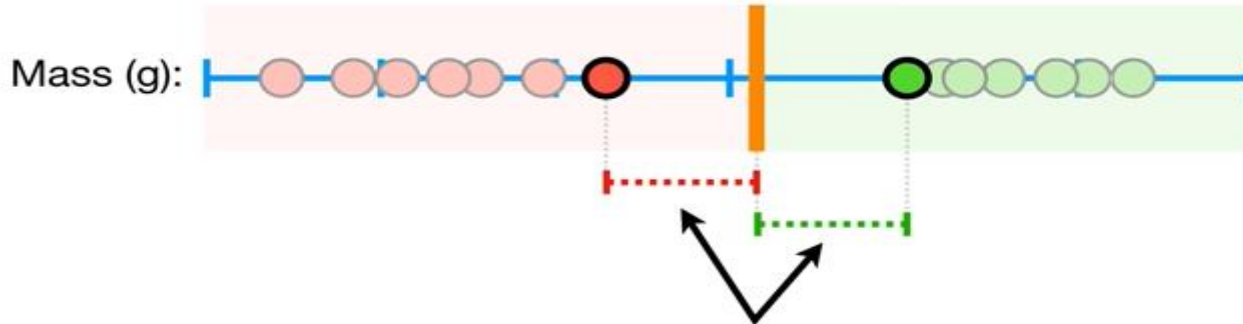


...we can focus on the observations on the edges of each cluster...



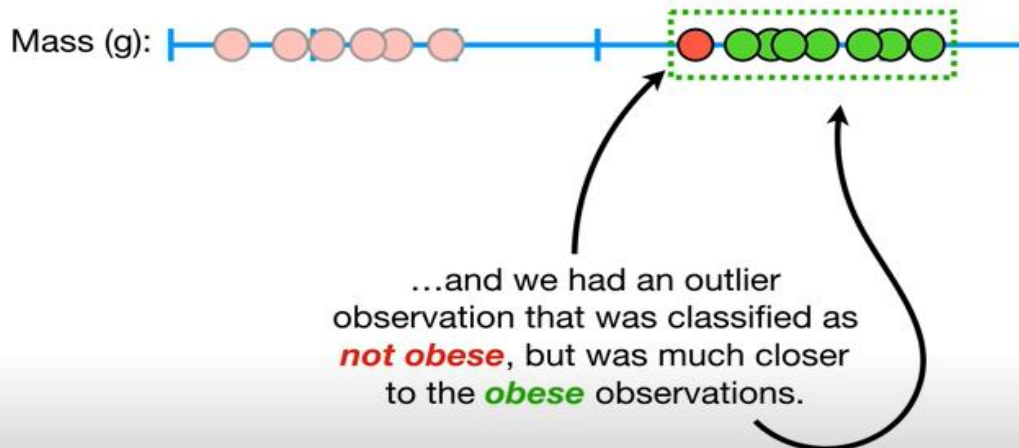
...and use the midpoint between them as the threshold.

SVM



The shortest distance between the observations and the threshold is called the **margin**.

Using margin with maximum distance for classification, called **Maximum Margin Classifier**

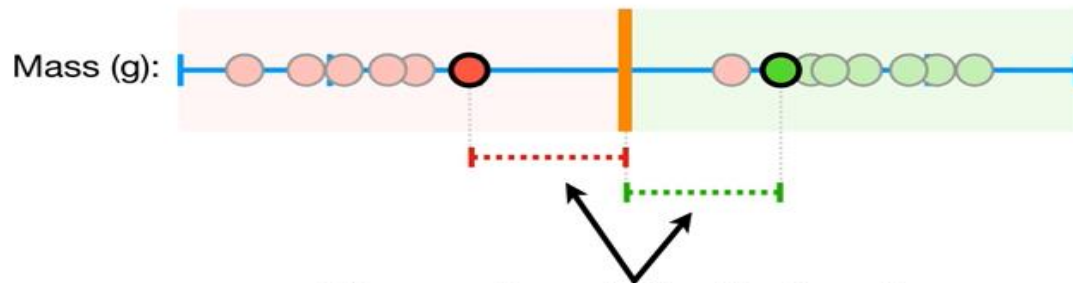


...and we had an outlier observation that was classified as **not obese**, but was much closer to the **obese** observations.

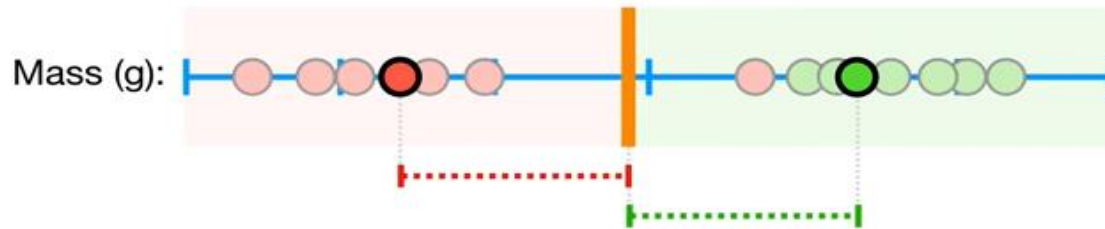
So **Maximal Margin Classifiers** are *super sensitive to outliers* in the training data and that makes them pretty lame.

SVM

Choose a threshold that allows misclassifications. This is an example of Bias-Variance Tradeoff.



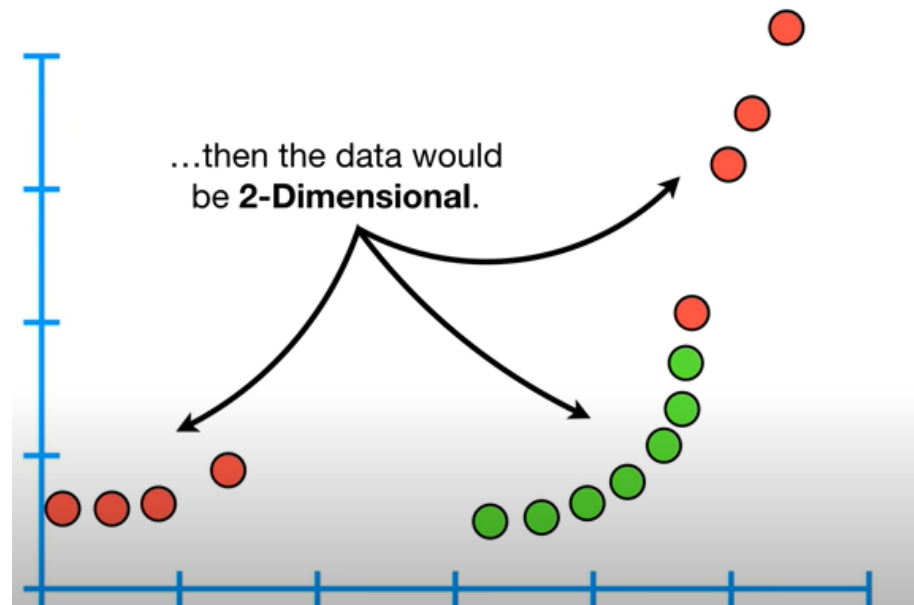
When we allow misclassifications, the distance between the observations and the threshold is called a **Soft Margin**.



The answer is simple: We use **Cross Validation** to determine how many misclassifications and observations to allow inside of the **Soft Margin** to get the best classification.

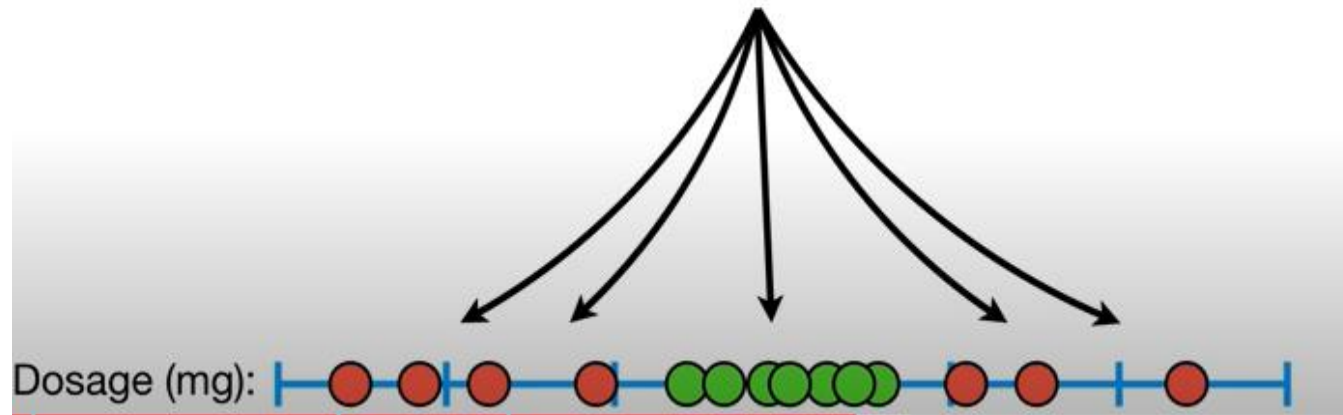
SVM

- When we use **Soft Margin** for classifier, we call it **Support Vector Classifiers**.
- For **Two dimensional** data, Margin will be a **Line**
- For **High dimensional** data, Margin will be a **plane** or **Hyperplane**



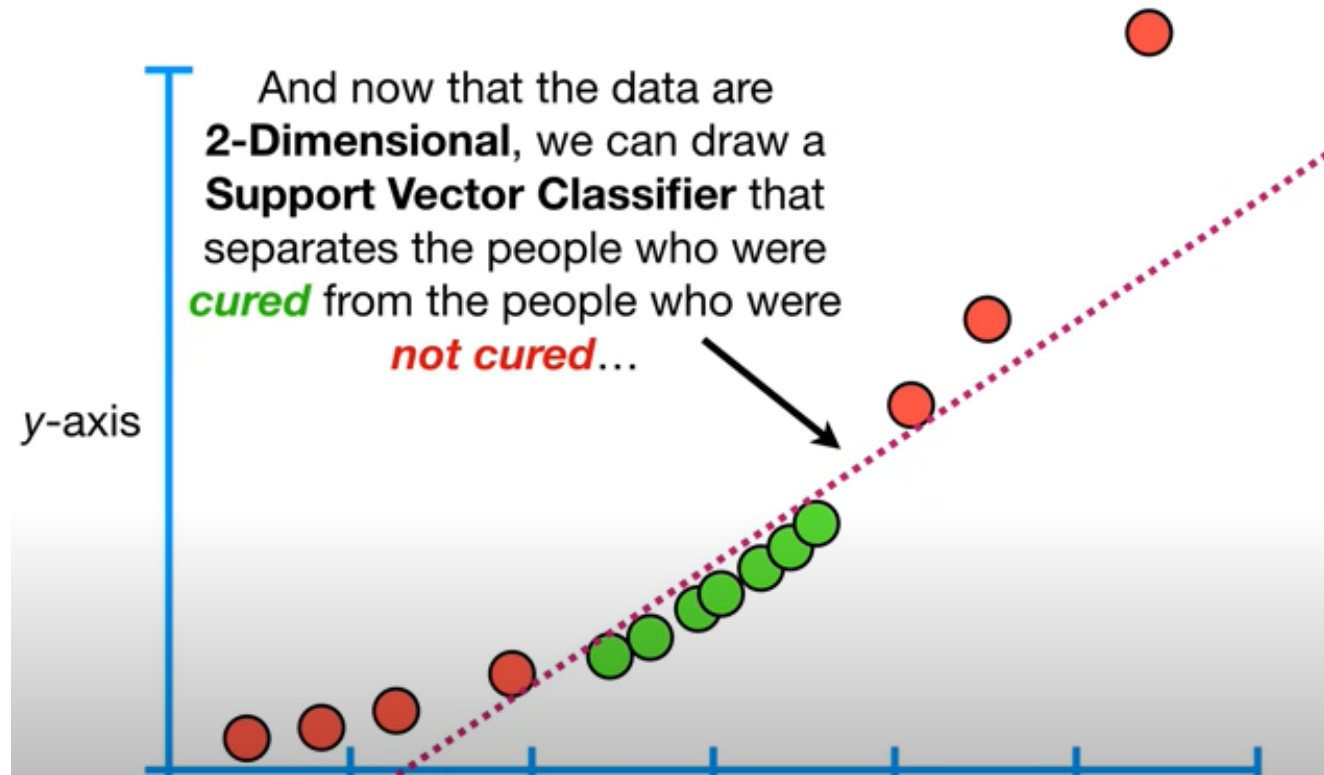
SVM

...but what if this was our training data and we had tons of overlap?



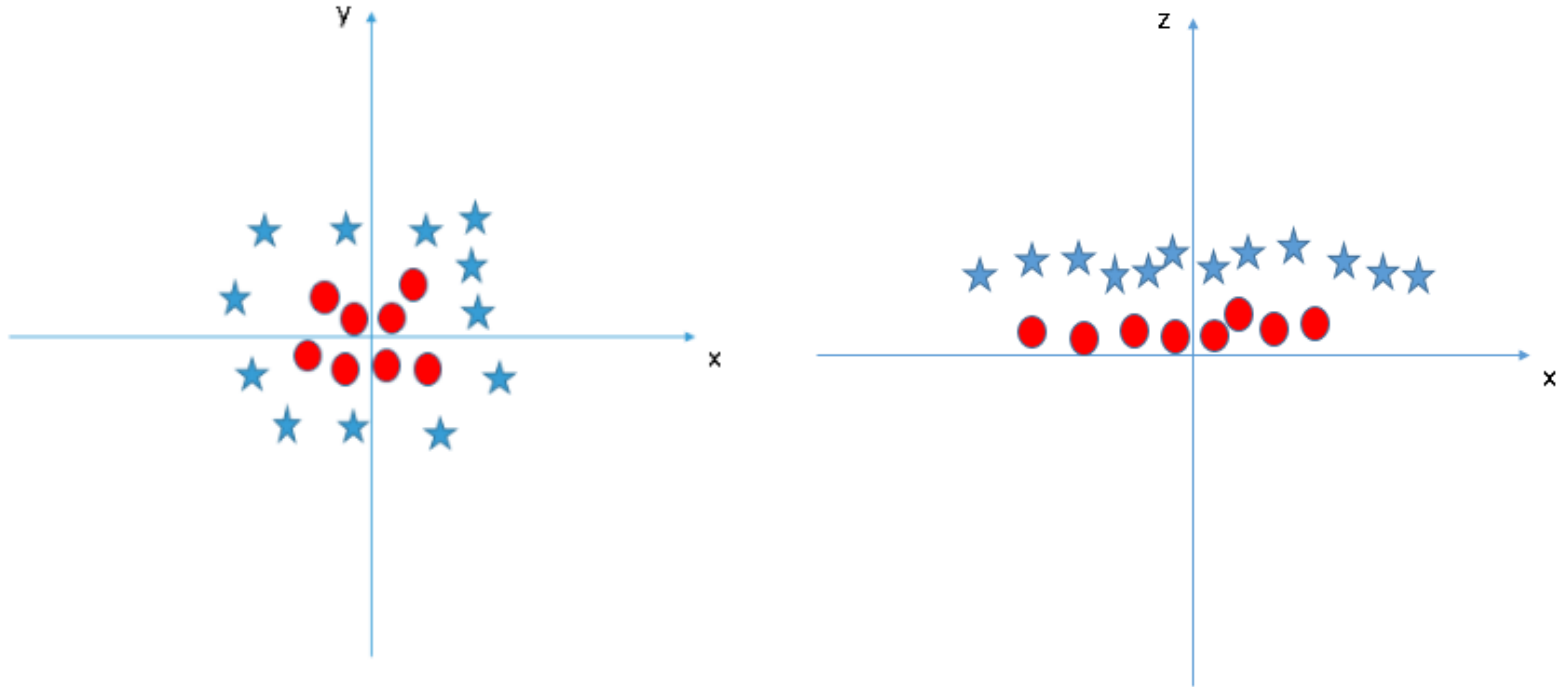
Support Vector Machine: Add an axis 'y' which will be Dosage²

Support Vector Machine



1. Start data in low dimension
2. Move data in higher dimension
3. Find Support Vector Classifier that separate data in higher space into different classes.

Support Vector Machine: Another Example



SVM solves this problem by introducing additional feature. Here, we will add a new feature $z = x^2 + y^2$. Now, let's plot the data points on axis x and z:

Support Vector Machine

How to determine higher dimensional space??

SVM uses **Kernal Functions** to find classifier in higher dimensions. SVM kernel is a function that takes low dimensional input space and transforms it to a higher dimensional space i.e. it converts not separable problem to separable problem. It is mostly useful in non-linear separation problem.

➤ Examples:

- Polynomial Kernal with degree d

- Radial Kernal or Radial Basis Function (RBF)