SVM Decision Boundary. Logistic Regression: A large C parameter tells the sum to try to classify all the example min C. Zi y'in. cost, (0^T. x'ii) + (1-y'i) cost, (0^T. x'ii)]+ Zi 0° => min $\frac{1}{2} = 0$, such that whenever $(y^{(i)} = 1, 0^{T}, x^{(i)} \ge 1)$ $U = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$ 11111 = length of vector u = Nu2+u2 ER Use SVM software package (e.g. liblinear, libsum, ...) to solve for parameters & Need to specify:

i) Choice of parameter C.

ii) Choice of kernel (similarity function).

Graussian kernel: $f: = \exp(-\frac{\pi x^{(i)}}{2\sigma^2})$, where $f(i) = \chi(i)$.

Read to choose $f(i) = \exp(-\frac{\pi x^{(i)}}{2\sigma^2})$, where $f(i) = \chi(i)$.

Polynomial kernel is $f(i) = \exp(-\frac{\pi x^{(i)}}{2\sigma^2})$ Folynomial kernel is $f(i) = \exp(-\frac{\pi x^{(i)}}{2\sigma^2})$ $f(i) = \exp(-\frac{\pi x^{(i)}$ i). if n is large, use logistic regression or sum without a kernel 11). it n is small, m is intermediate, use SUM with Gaussian kernel iii). If nis small, mis large, evente add more features, then do i).

Support Vector Machine

Clustering K-Mean algorithm: 1. Randomly initialize K cluster centroids u, uz, ..., UKER. 2. Repeat \$ for i=1 to m (m is the # of examples) C(i) = index (from 1 to K) of cluster centroid closest to $\chi(i)$. cluster assignment step. for k=1 to k: uk = anerage of points assigned to cluster k Assigned cluster centroid. min J(c(1),...,c(m), u1,...,uk) = = = = = [1x(i)-Uciv)[Optimization Objective: choose the number of k: Elbow method:

& Elbow.