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- a) for class y=0 E[x1] = (1+1+2+3+3)/5 = 2 E[x2] = (1+1+2+2+3)/5 = 1.8 cov(x1,x2) = E[x1x2] E[x1]E[x2] = (3+6+6+1+2)/5 3.6 = 0 for class y=1 E[x1] = (1+2+4+5+5)/5 = 3.4 E[x2] = (4+5+6+6+7)/5 = 5.6 cov(x1,x2) = E[x1x2] E[x1]E[x2] = (30+24+20+10+7)/5 19.04 = 18.2-19.04 = -0.84 b) xT = (3.5,2) sigma = covariance matrix for y=0 var(x1) = E[x1^2] E[x1]^2 = (1+1+4+9+9)/5 4 = 0.8var(x2) = (1+1+4+4+4+9)/5 3.6 = 0.2 fory = 1 var(x1) = (1+4+16+25+25)/5 11.56 = 14.2-11.56 = 2.64 var(x2) = <math>(16+25+36+36+49)/5-31.36 = 32.4-31.36 = 1.04 sigma fory = 0(0.8,0)(0,0.2) sigma fory = 1(2.64,-0.84)(-0.84,1.04) mu = vector of expected values
- c) LDA assumes that the classes have different means and shared variance, while with QDA each class can have a different variance
 - d) ???
- e) LDA is a much less flexible classifier than QDA, therefore LDA usually makes better predictions when there are relatively few training observations and reducing variance is crucial. QDA can be used with a bigger sample size, when the variance of the classifier is not a huge concern.

number of features??