

## Fast K-NN classification (extra pruning tricks)

*K-nearest-neighbor classification without finding the K-NN*

The Auton Lab  
Carnegie Mellon University

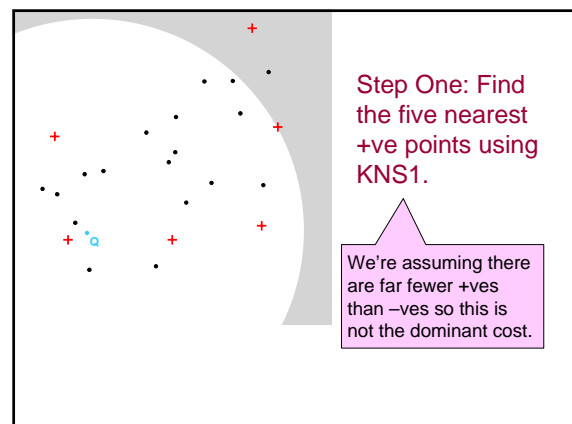
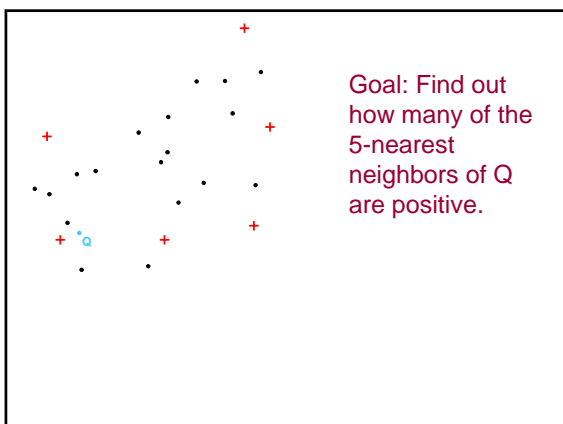
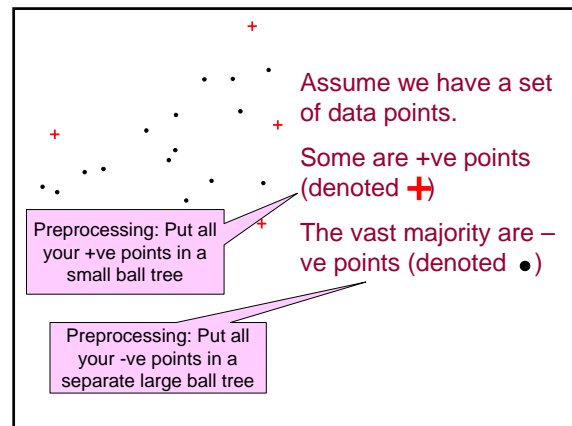
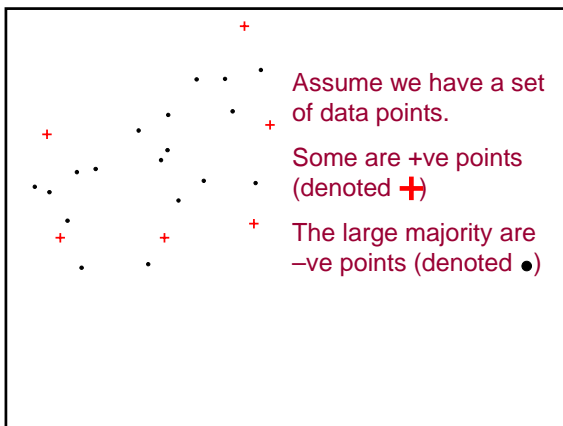


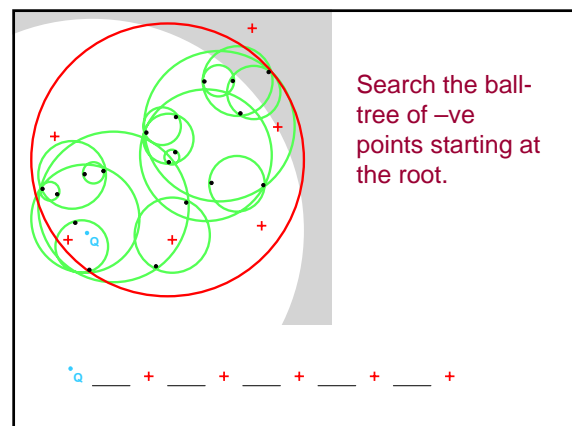
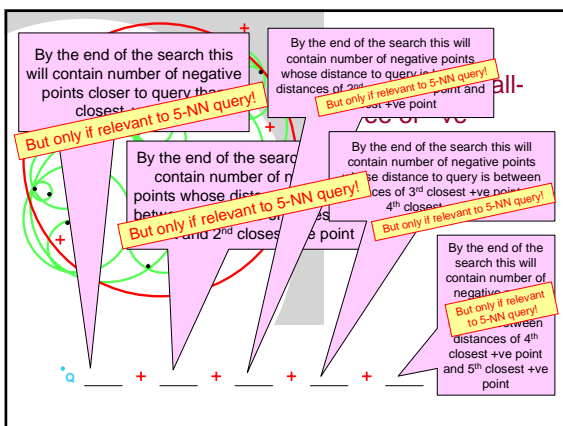
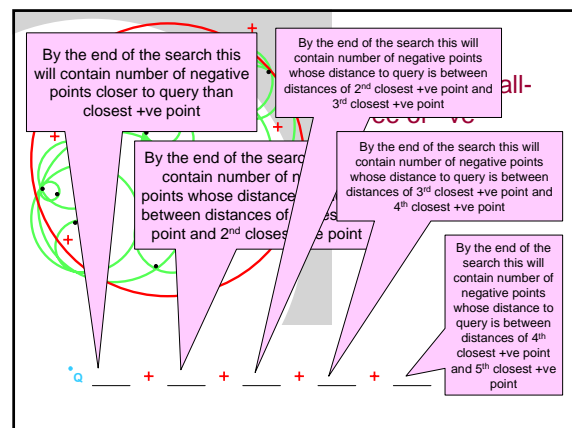
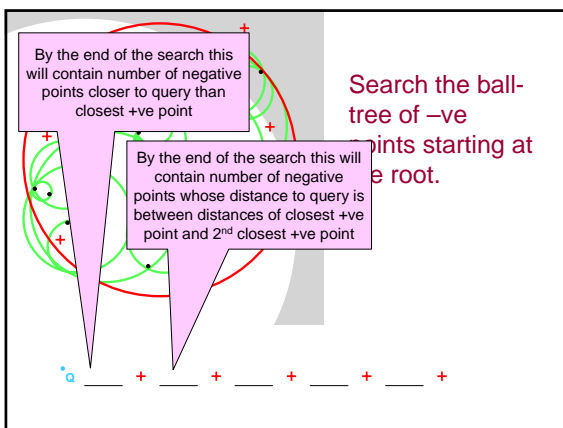
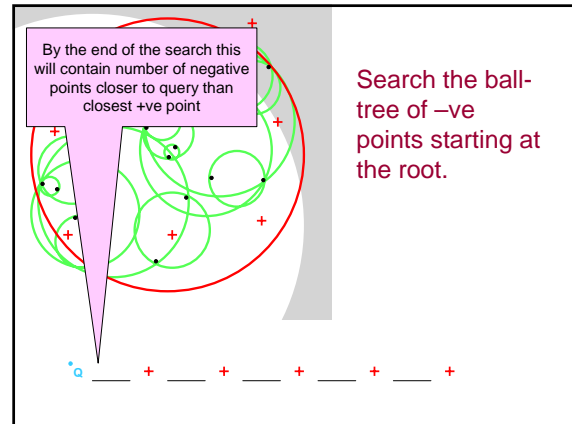
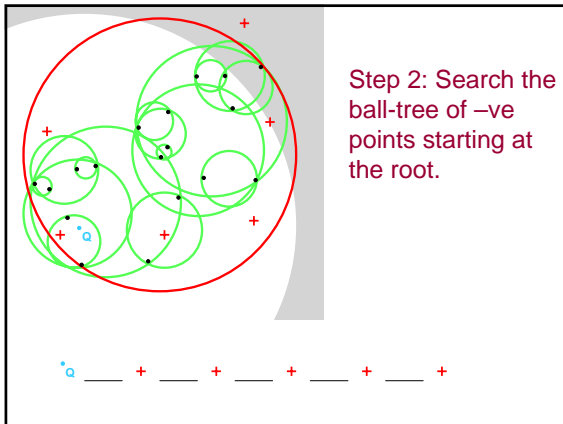
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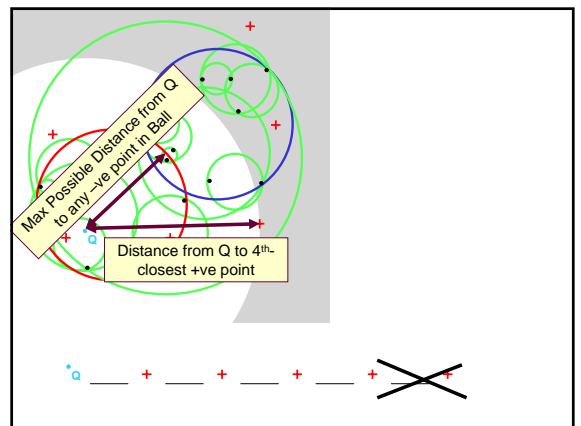
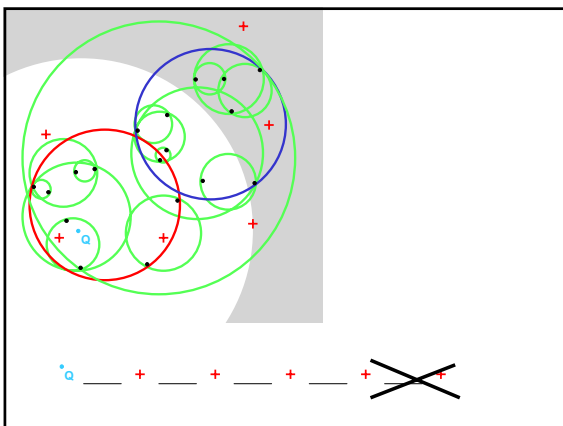
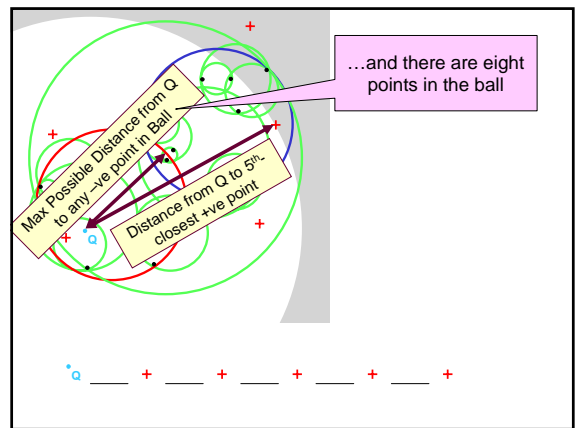
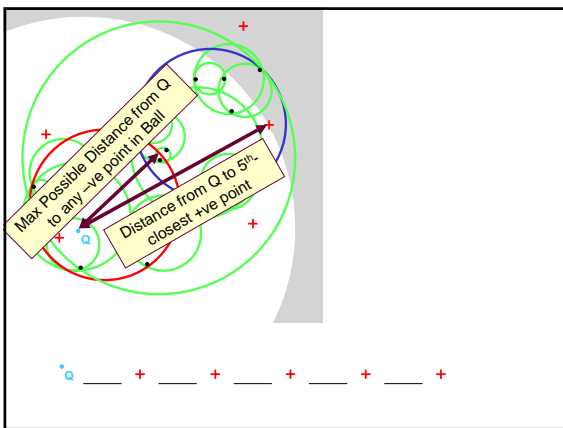
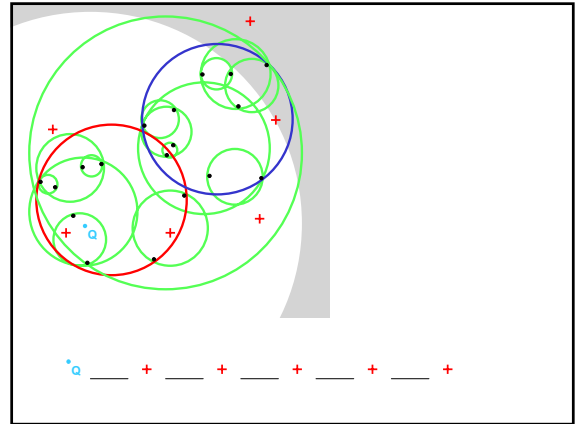
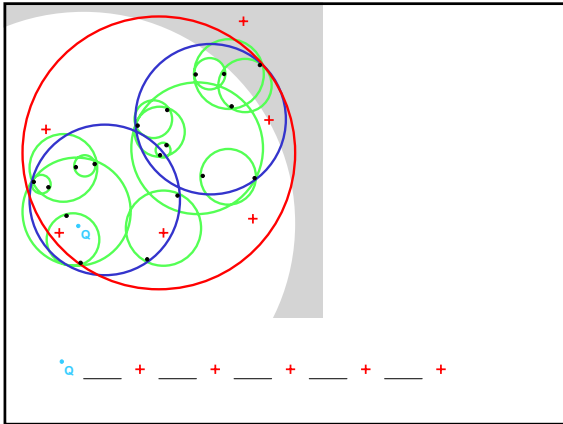
## KNS2

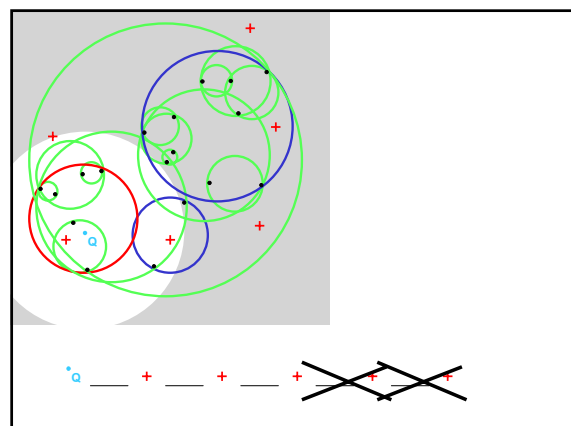
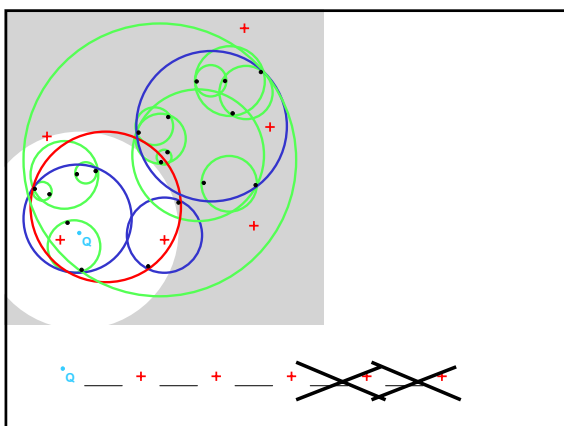
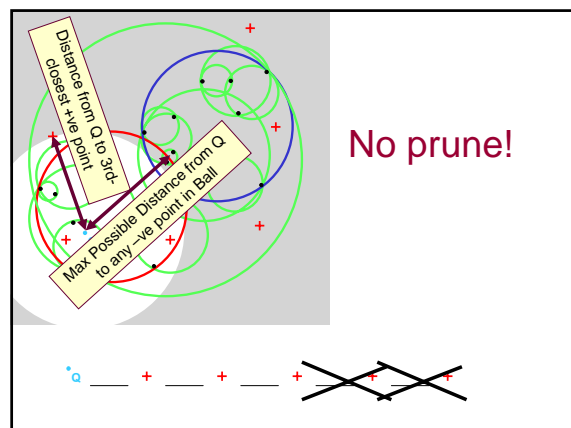
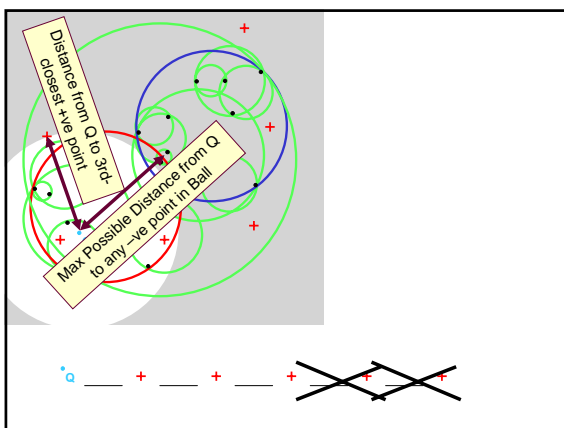
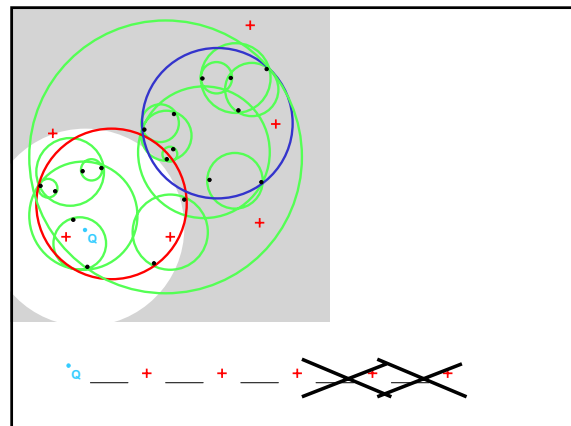
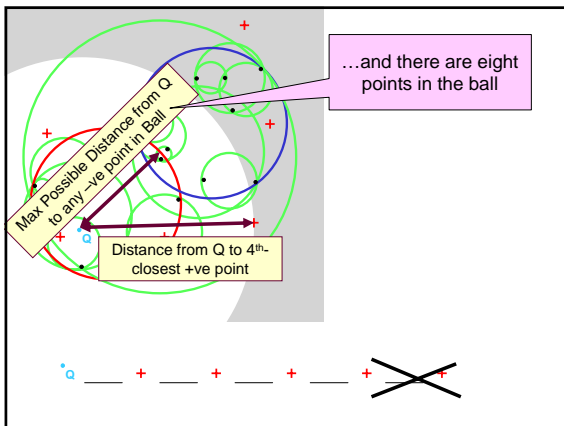
- Assume binary output
- Assume positive class is much less frequent than negative class
- Assume we want more than a “positive/negative” prediction: we want to know exactly how many of the K-NN are from the +ve class

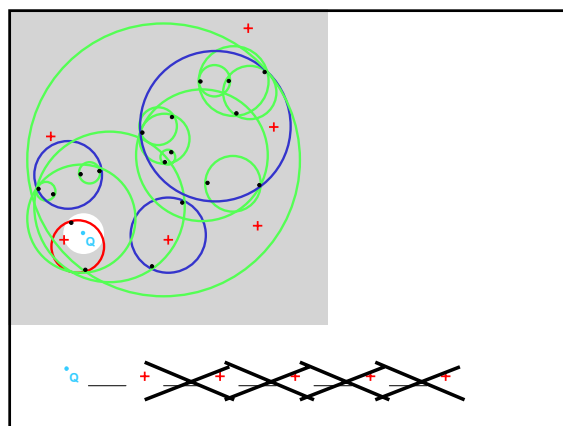
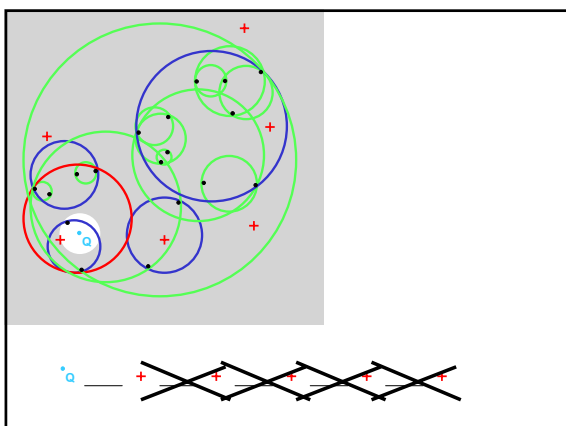
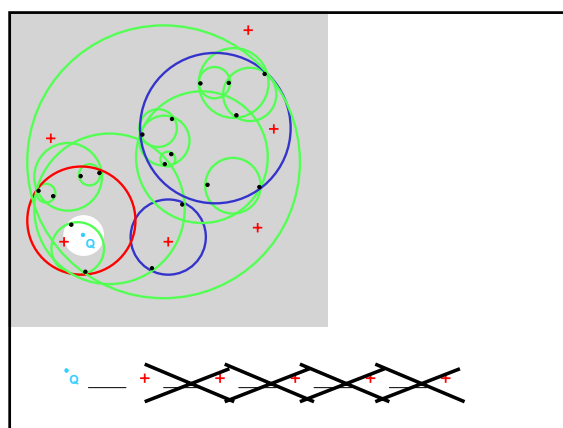
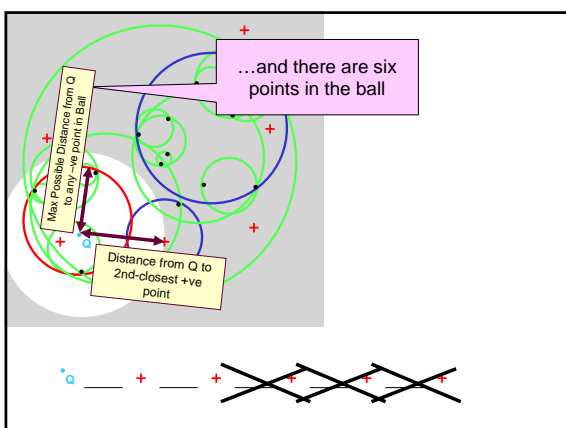
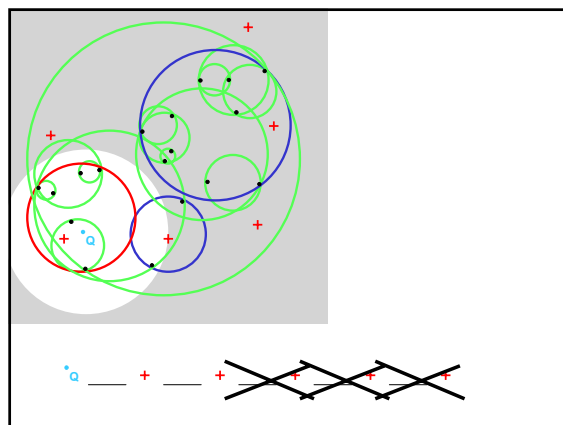
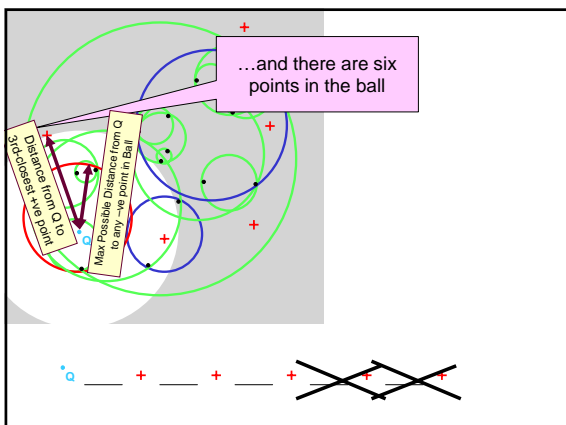
KNS2 does this without finding the K-NN

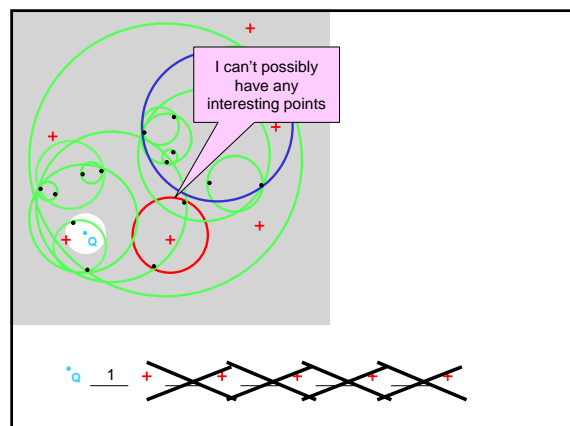
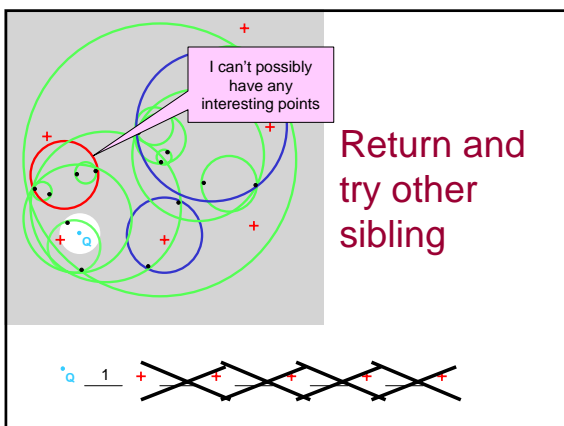
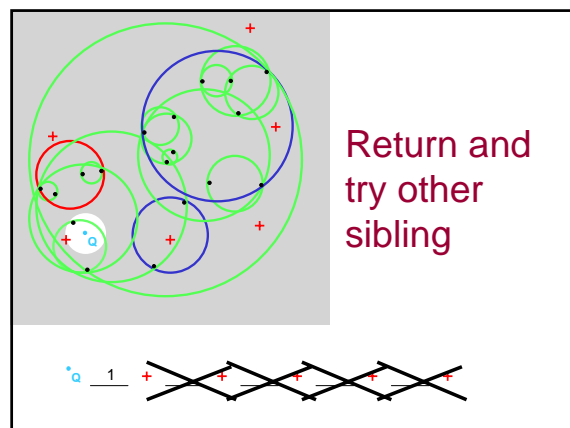
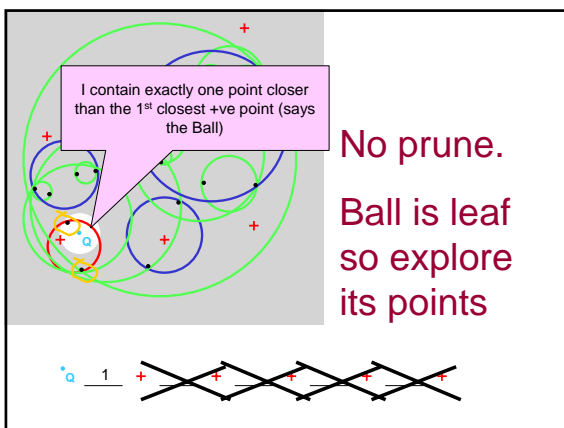
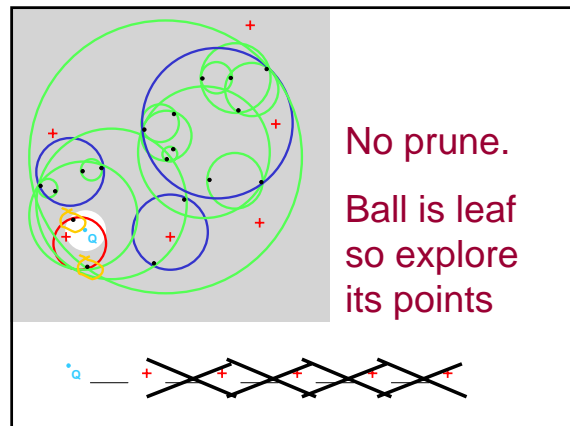
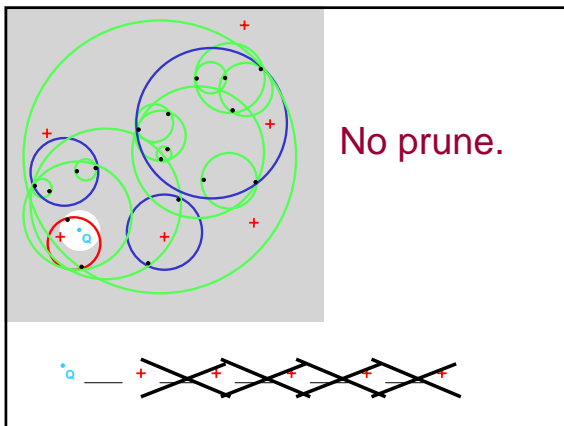


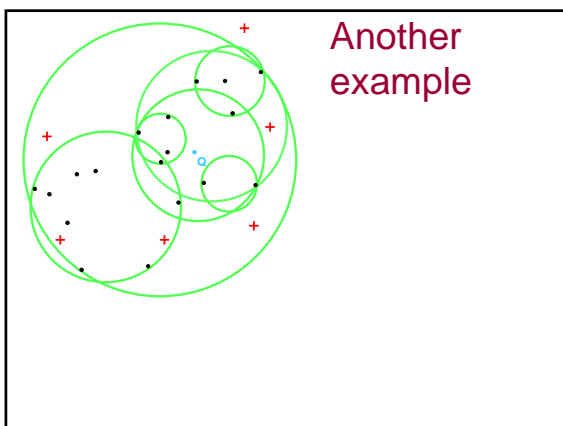
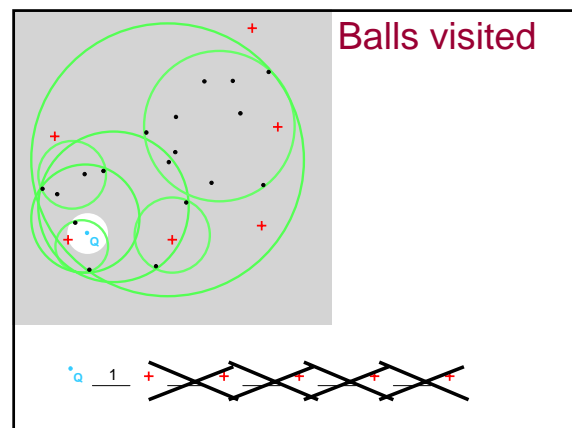
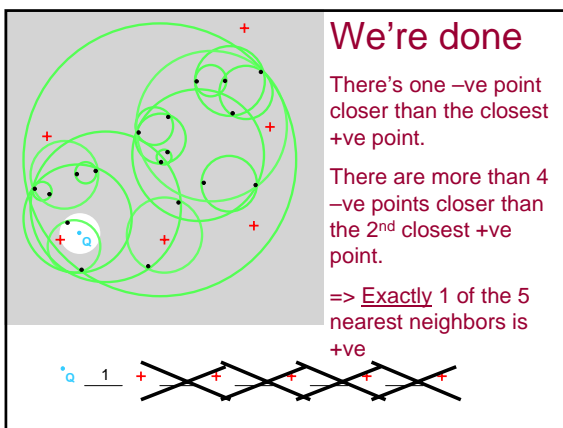
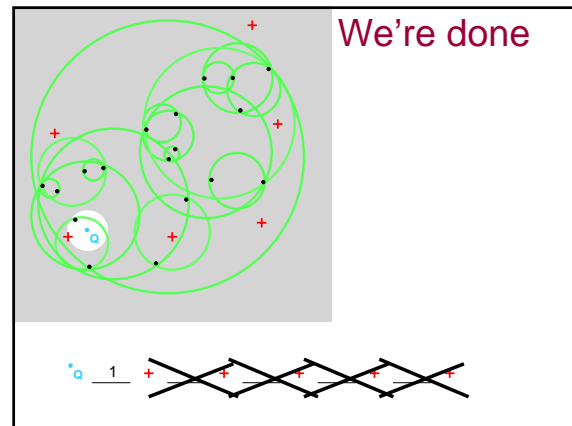
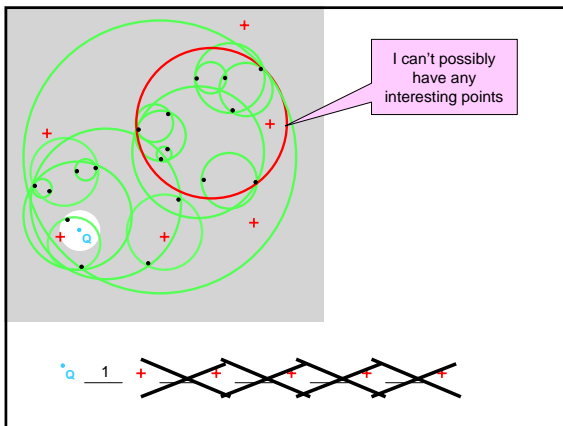












**Conclusion**

- It is often possible to exploit specific features of your statistical problem to give more pruning opportunities than were in the familiar geometrical problem.
- Here we worked with the special case of binary skewed classification: other cases have also been developed.