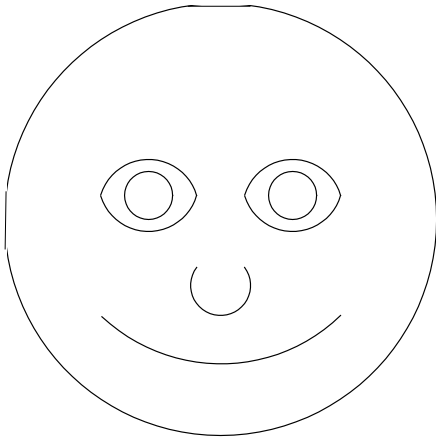


HW #10

1. **Model Matching (10 pts):** This problem examines the detection of faces by model matching. A face is modeled as a collection of circles and circular arcs, as so:



The basic idea is that face detection begins by finding circles and circular arcs in the image, followed by matching against a stored model. Be sure to give short justifications for each answer below.

- How should the model be represented?
 - How should images be processed to detect the features?
 - How should matching be performed?
 - Is your answer to c. invariant to translation, rotation, and scale?
 - Can it handle partial occlusion? If so, how? If not, suggest an extension to your scheme that can.
2. **Interpretation Tree (5 pts):** We have a choice of matching detected image elements (edges) to the model or model elements to the object. Let E be the set of detected image edges and M the set of model edges. In the first case, matching image edge to model edges, we generate a tree of depth $|E|$ and breadth $|M|$ with tree size $|M|^{|E|}$. In the case of matching the model to image, we generate a tree of size $|E|^{|M|}$. We expect many more image elements than model elements – there may be many candidate image edges in a cluttered scene vs. a small number of model edges.
- Which approach is preferable, matching image edges to model or model to image edges? You might consider the case where there are 12 image edges and 5 model edges, for example.

- b. One advantage to using the interpretation tree approach is that it is possible to match an unknown object in the image to a model even if the object is partially occluded. We do this by allowing an object element to match a “null element” in the model. Does this change your answer to part a.? How and why? Or why not?

3. **Binary Image Matching (2 pts):** Let I_1 and I_2 be binary images. Show that

$$|I_1 - I_2|^2 = \sum \# \text{ of pixels where } I_1 \neq I_2$$

Where $|I|^2 = \sum i_{jk}^2$ is the sum of all (pixels squared) in I .

4. **Classification (5 pts):** Suppose we have a 2-class classification problem with class means $\vec{\mu}_A$ and $\vec{\mu}_B$. Assuming that both classes are equally likely, show that the Nearest Mean classifier decision boundary is the hyper-plane perpendicular to, and midway along, the line segment connecting $\vec{\mu}_A$ to $\vec{\mu}_B$. You do not need to assume any particular distribution for classes A and B .