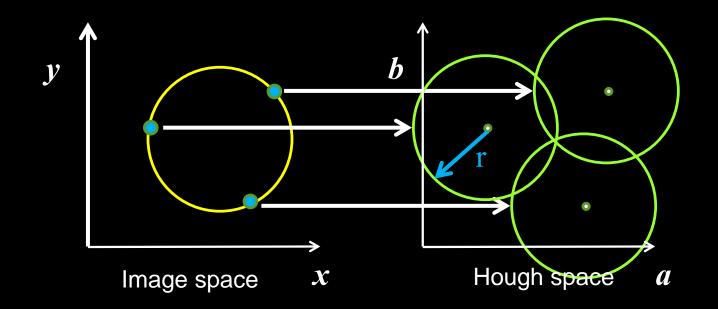
CS4495/6495 Introduction to Computer Vision

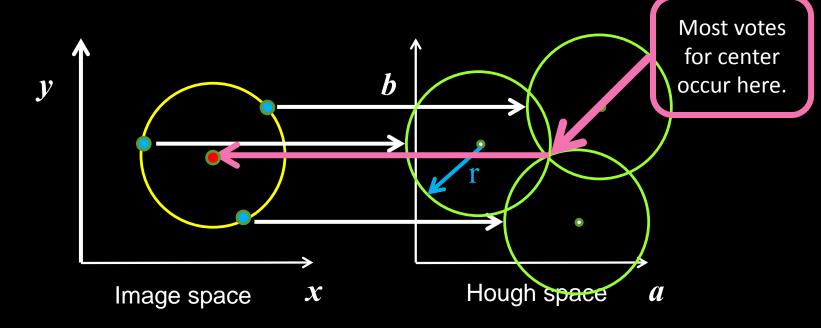
2B-L2 Hough transform: Circles

- Circle: center (a,b) and radius $r(x_i a)^2 + (y_i b)^2 = r^2$
- For a fixed radius r, unknown gradient direction:



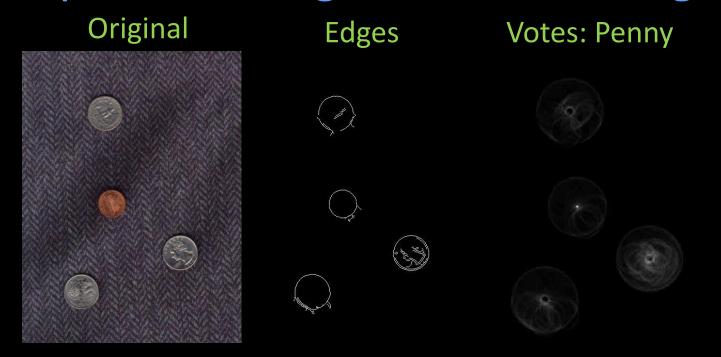
• Circle: center (a,b) and radius $r(x_i - a)^2 + (y_i - b)^2 = r^2$

For a fixed radius r, unknown gradient direction:

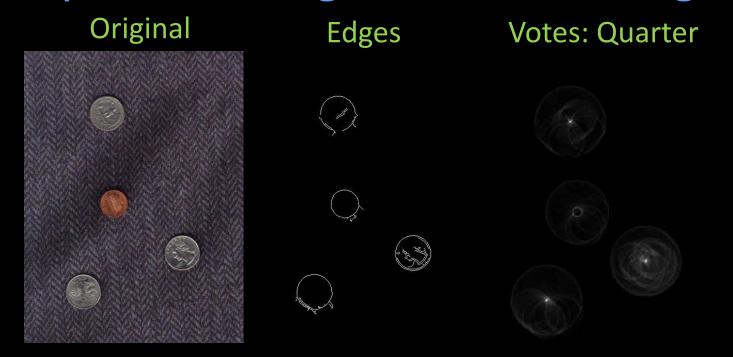




Crosshair indicates results of Hough transform; bounding box found via motion differencing.

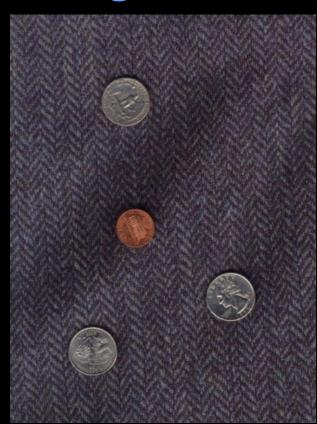


Note: a different Hough transform (with separate accumulators) was used for each circle radius (quarters vs. penny).



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Original

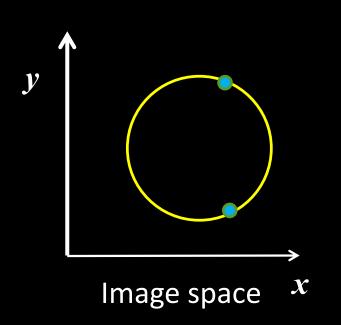


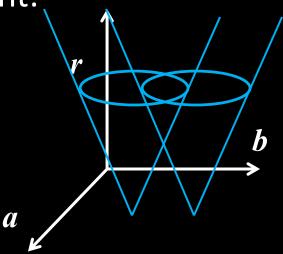
Combined detections



• Circle: center (a,b) and radius $r(x_i - a)^2 + (y_i - b)^2 = r^2$

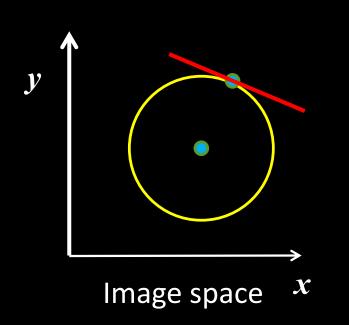
• For *unknown* radius r, no gradient:

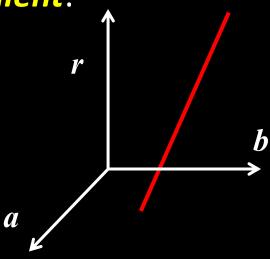




Hough space

- Circle: center (a,b) and radius $r(x_i a)^2 + (y_i b)^2 = r^2$
- For *unknown* radius r, with *gradient*:





Hough space

end

```
For every edge pixel (x,y):
        For each possible radius value r:
          For each possible gradient direction \theta:
3.
            %% or use estimated gradient
            a = x - r \cos(\theta)
4.
              b = y + r \sin(\theta)
5.
               H[a,b,r] += 1
6.
          end
7.
        end
8.
```

Voting: practical tips

- Minimize irrelevant tokens first (take edge points with significant gradient magnitude)
- Choose a good grid / discretization:
 - Too coarse: large votes obtained when too many different lines correspond to a single bucket
 - Too fine: miss lines because some points that are not exactly collinear cast votes for different buckets

Voting: practical tips

- Vote for neighboring bins (like smoothing in accumulator array)
- Utilize direction of edge to reduce free parameters by 1
- To read back which points voted for "winning" peaks, keep tags on the votes

Parameterized Hough transform: pros and cons

<u>Pros</u>

- All points are processed independently, so can cope with occlusion
- Some robustness to noise: noise points unlikely to contribute consistently to any single bin
- Can detect multiple instances of a model in a single pass

Parameterized Hough transform: pros and cons

Cons

- Complexity of search time increases exponentially with the number of model parameters
- Non-target shapes can produce spurious peaks in parameter space
- Quantization: hard to pick a good grid size