

Group 24
God of Gamblers



Problem Definition

- Identifying playing cards (4 suits × 13 ranks) within a scene
 - Dark background with white cards
 - Cards are not covered
 - These assumptions may be relaxed

Preprocessing

- Obtain properly oriented card images to build a database
- Extracting the card from the (dark) background.
- Locating the perimeters of the card
- Affine transformation
- Filtering the results

Extracting the card from the (dark) background

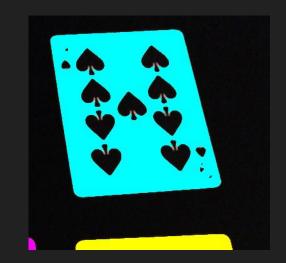
- Grayscale image
- Gaussian smoothing
 - Reduce Gaussian noise while maintaining details
- Thresholding
 - o **127**
 - Otsu's method
 - Perform a 2-clustering on all pixels, with centers 0 and 255

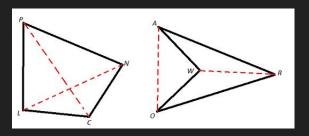




Locating the perimeters

- Connected components
 - In the ideal case, white (object) pixels are separated into components by black (background) pixels
 - A card is simply a connected component
- 2-D Convex hull
 - By definition, a convex hull gives the tightest bounding convex polygon of a set of points
 - A perspectively projected rectangle must be a convex quadrilateral
 - Proof by contradiction: suppose it becomes concave ...
- Polygon approximation
 - Reduce the number of corners to 4





Affine transformation

- Converting a perspectively projected rectangular patch to its upright position
 - At this stage we permit a rotated card image there are at most 4 cases to check, in the recognition stage
 - Now we can create our training database





Filtering the results

- Purpose
 - To remove some false positives
- Shape
 - o Inertia: how elongated an object is
 - In our program, we limit the aspect ratio of the white patch to between 1 : 0.6 and 0.6 : 1
- Blob detection
 - A card should contain some symbols, which are dark pixels within a white background
 - If a white region does not contain any blobs, it does not qualify as a card
 - Simply check the min/max coordinates
- Components fully inscribed within another



Limitations

- Highly dependent on the thresholding operation
 - If it fails to remove the background, all the remaining computations would fail
 - Runs into similar problems if a card is partially covered
 - Image segmentation is non-trivial
 - Vs. SIFT/SURF: these patented algorithms perform much better as they are agnostic to pixels othe interest (ROI)



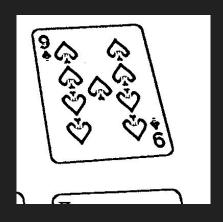




Limitations

Complexity

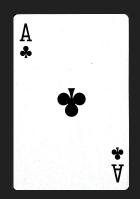
- Current implementation finds the convex hull of all white pixels - Ω(n log n) worst-case running time in SciPy implementation
 - n is the number of white pixels in one connected component
 - n can be reduced by operating on the negative Harris corner response image, for instance
 - Or by performing adaptive (local) thresholding
 - Flood fill is O(n)



Classifying rank and suit

- Problem
- Identifying the rank and suit of a thresholded, upright-positioned rectangular playing card with no background







- Solution
- Extract feature from image
 - Position of contour point from xaxis and y-axis
 - 10×10 histogram of highlighted pixel from x-axis and y-axis
 - Wavelet coefficients
- SIFT algorithm

Extract feature from image

- Identify contours using first principles
- Algorithm:
- Find contours
- Group each contour together with its first layer of children into one
 - Due to the presence of holes in rank "A", "4","6", "8", "9", "Q"
- Crop out for further analysis



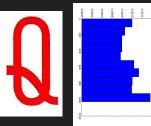




Extract feature: Compare contour position

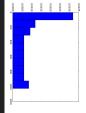
- Compressed cropped area into two list of data
 - Histogram showing the probability distribution of contour points in certain range of x-position
 - Histogram showing the probability distribution of contour points in certain range of y-position
- Compared the histograms of template and test image
 - By correlation
 - By chi-squared
- Limitation
 - Non-rotational invariant
 - Requires high resolution
 - Weak in differentiating object with different "thickness", e.g. heart and spade in low resolution





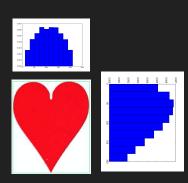


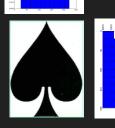




Extract feature: Compare highlighted pixel

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Distortions

- Wavelet compression
 - Works well for digitized signals
 - Separates an image into approximation and detail (c.f. low-pass and high-pass filters in signal processing)
 - Reconstructing an approximate image from less coefficients than the raw image
 - Thereby allowing for some minor distortions
 - What exactly has been compressed?
 - Signals (images) in real life do not tile indefinitely
- Limitation
 - Non-rotational invariant
 - Low separation power among cards with many edges J,
 Q, K





SIFT (Scale-Invariant Feature Transform)

- Another efficient feature extraction algorithm
- First proposed by David Lowe, in 1999
- To find out key-points (with descriptor vectors) of an image
- Robustness
 - o Scale
 - Rotation
 - Lighting
 - Noises
- Recognition rate: > 90%

SIFT (Scale-Invariant Feature Transform)

1. Scale space extrema detection

- Resize image into N different scales (usually N=5)
- Use Difference of Gaussian (DoG) filter
- Find local extrema as potential key-points: {x, y, scale}

Keypoint filtering

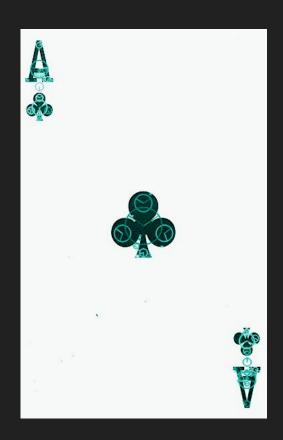
- Using thresholding, to eliminate low-contrast keypoints
- Using edge detector, to eliminate edged keypoints

3. Orientation

- Using Gradient Histogram in 36 directions, to find maxima
- Get keypoints list: {x, y, scale, orientation}

4. Matching

- Identifying nearest neighbours in key-points of two images
- Use Brute-Force matching algorithm



Applications

- Play annotations
- Strategy suggestions