



discriminate between silhouettes of model cars, vans and buses viewed from constrained elevation but all angles of rotation.

The rule tree classification performance compared favourably to MDC (Minimum Distance Classifier) and k-NN (k-Nearest Neighbour) statistical classifiers in terms of both error rate and computational efficiency. An investigation of these rule trees generated by example indicated that the tree structure was heavily influenced by the orientation of the objects, and grouped similar object views into single decisions.

#### DESCRIPTION

The features were extracted from the silhouettes by the HIPS (Hierarchical Image Processing System) extension BINATTS, which extracts a combination of scale independent features utilising both classical moments based measures such as scaled variance, skewness and kurtosis about the major/minor axes and heuristic measures such as hollows, circularity, rectangularity and compactness.

Four "Corgie" model vehicles were used for the experiment: a double decker bus, Cheverolet van, Saab 9000 and an Opel Manta

400.

This particular combination of vehicles was chosen with the expectation that the bus, van and either one of the cars would be readily distinguishable, but it would be more difficult to distinguish between the cars.

The images were acquired by a camera looking downwards at the model vehicle from a fixed angle of elevation (34.2 degrees to the horizontal). The vehicles were placed on a diffuse backlit surface (lightbox). The vehicles were painted matte black to minimise highlights. The images were captured using a CRS4000 framestore connected to a vax 750. All images were captured with a spatial resolution of 128x128 pixels quantised to 64 greylevels. These images were thresholded to produce binary vehicle

silhouettes,

negated (to comply with the processing requirements of BINATTS) and thereafter subjected to shrink-expand-expand-shrink HIPS modules to remove "salt and pepper" image noise.

The vehicles were rotated and their angle of orientation was measured

using a radial graticule beneath the vehicle. 0 and 180 degrees corresponded to "head on" and "rear" views respectively while 90

and

270 corresponded to profiles in opposite directions. Two sets of 60 images, each set covering a full 360 degree rotation, were

captured

for each vehicle. The vehicle was rotated by a fixed angle between images. These datasets are known as e2 and e3 respectively.

A further two sets of images, e4 and e5, were captured with the camera

at elevations of 37.5 degs and 30.8 degs respectively. These sets also contain 60 images per vehicle apart from e4.van which contains only 46 owing to the difficulty of containing the van in the image at some orientations.

## ATTRIBUTES

COMPACTNESS  $(\text{average perim})^2/\text{area}$

CIRCULARITY  $(\text{average radius})^2/\text{area}$

DISTANCE CIRCULARITY  $\text{area}/(\text{av.distance from border})^2$

RADIUS RATIO  $(\text{max.rad}-\text{min.rad})/\text{av.radius}$

PR.AXIS ASPECT RATIO  $(\text{minor axis})/(\text{major axis})$

MAX.LENGTH ASPECT RATIO  $(\text{length perp. max length})/(\text{max length})$

SCATTER RATIO  $(\text{inertia about minor axis})/(\text{inertia about major axis})$

ELONGATEDNESS  $\text{area}/(\text{shrink width})^2$

PR.AXIS RECTANGULARITY  $\text{area}/(\text{pr.axis length}*\text{pr.axis width})$

MAX.LENGTH RECTANGULARITY  $\text{area}/(\text{max.length}*\text{length perp. to this})$

SCALED VARIANCE (2nd order moment about minor axis)/area  
ALONG MAJOR AXIS

SCALED VARIANCE (2nd order moment about major axis)/area  
ALONG MINOR AXIS

SCALED RADIUS OF GYRATION  $(\text{mavar}+\text{mivar})/\text{area}$

SKEWNESS ABOUT (3rd order moment about major axis)/ $\sigma_{\text{min}}^3$   
MAJOR AXIS

SKEWNESS ABOUT (3rd order moment about minor axis)/ $\sigma_{\text{maj}}^3$   
MINOR AXIS

KURTOSIS ABOUT (4th order moment about major axis)/ $\sigma_{\text{min}}^4$   
MINOR AXIS

KURTOSIS ABOUT (4th order moment about minor axis)/ $\sigma_{\text{maj}}^4$   
MAJOR AXIS

HOLLOWS RATIO  $(\text{area of hollows})/(\text{area of bounding polygon})$

Where  $\sigma_{\text{maj}}^2$  is the variance along the major axis and  $\sigma_{\text{min}}^2$  is the variance along the minor axis, and

$\text{area of hollows} = \text{area of bounding poly} - \text{area of object}$

The area of the bounding polygon is found as a side result of the computation to find the maximum length. Each individual length computation yields a pair of calipers to the object orientated at every 5 degrees. The object is propagated into

an image containing the union of these calipers to obtain an image of the bounding polygon.

#### NUMBER OF CLASSES

4 OPEL, SAAB, BUS, VAN

#### NUMBER OF EXAMPLES

Total no. = 946

No. in each class

opel	240
saab	240
bus	240
van	226

100 examples are being kept by Strathclyde for validation.  
So StatLog partners will receive 846 examples.

#### NUMBER OF ATTRIBUTES

No. of atts. = 18

#### BIBLIOGRAPHY

Turing Institute Research Memorandum TIRM-87-018 "Vehicle  
Recognition Using Rule Based Methods" by Siebert, JP (March 1987)