



also shadows of buildings around these roads lie on them. In a low resolution image the size of pixel could be bigger than the width of road so the lower the image resolution the harder to detect the roads (Figure 2).

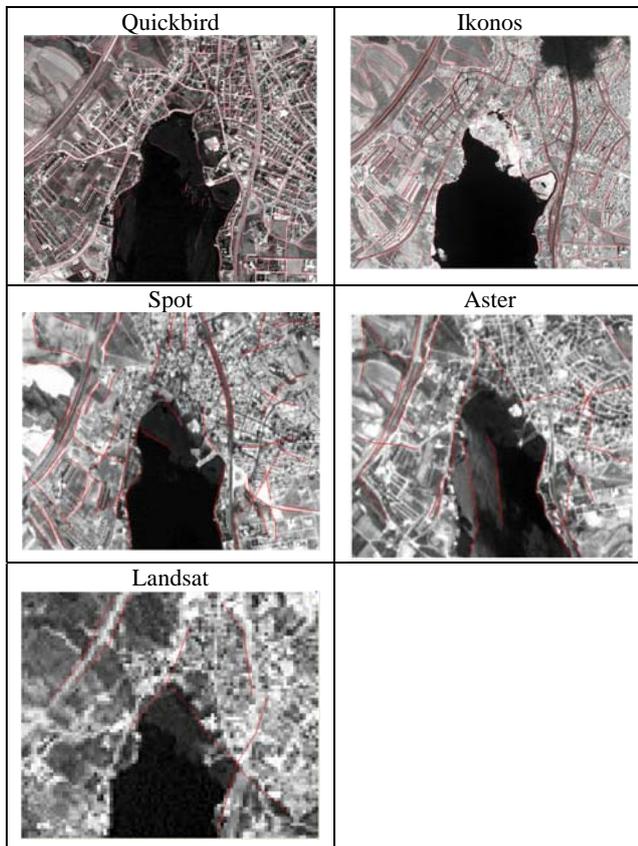


Figure 2: Roads that extracted by using automated methods.

## 2.2. Semi-Automated Methods

### Filtering

Spatial frequency is described as the number of variations among pixels' values in a specific region over the raster dataset. If this variation is low that image can be described as low frequency image, otherwise if variation is high the image described as high frequency image.

High pass filters are used for increasing the spatial frequency of images while Low pass filters are used for reducing or suppressing the spatial frequency of images. In filtering processes each pixel evaluated with particular number of its neighbors pixels and depends on weight the new value of each pixel is computed and then assigned to each one.

In this study in order to detect the roads edge detection filters which are the types of high pass filters (Prewitt and Sobel) are conducted to each satellite images having different spatial resolution. In this way spatial frequency of each image was increased so the roads were highlighted and then extracted from each image (Figure 3). According to this processes' results, in high resolution images such as Quickbird and Ikonos the main and secondary roads are detected more accurately, however in low resolution images (Spot, Aster and Landsat-ETM) the details are reduced and detection of roads become more difficult therefore at some region only the main roads could be detected.

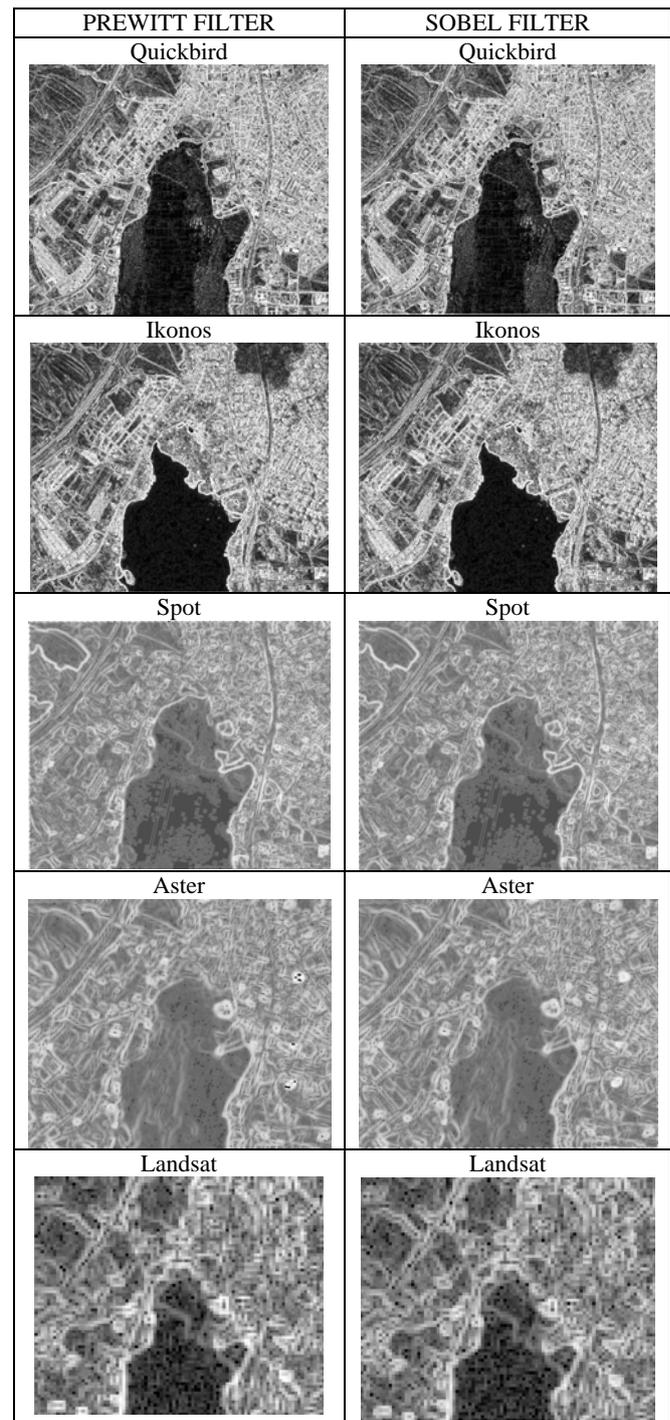


Figure 3: High pass filtered (Prewitt and Sobel) images.

### Classification

Classification can be described as, grouping image pixels into categories or classes to produce a thematic representation. Classification can be used in thematic maps or can be further incorporated into digital analysis. It can be performed on single or multiple image channels to separate areas according to their different scattering or spectral characteristics. Digital image classification procedures are differentiated as being either supervised or unsupervised.

In this study Maximum Likelihood classification technique was applied four images (Quickbird, Ikonos, Aster, and Landsat)

that have different resolution. Six classes were defined; and for each class training data were collected and images were classified. Then classified images were reclassified as two classes; road and nonroad by merging classes other than road and results were compared. Although the classified data of high resolution images produced better results, several problems experienced;

When the pavement of road surface changes (e.g. from asphalt to concrete) the contrast changes so the same road could be assigned different cluster label.

Due to very high resolution, vehicle, overpass, and other objects on a road could cause misclassification for that particular part. Other objects such as house roofs, which have same spectral property as roads, are misclassified into road clusters.

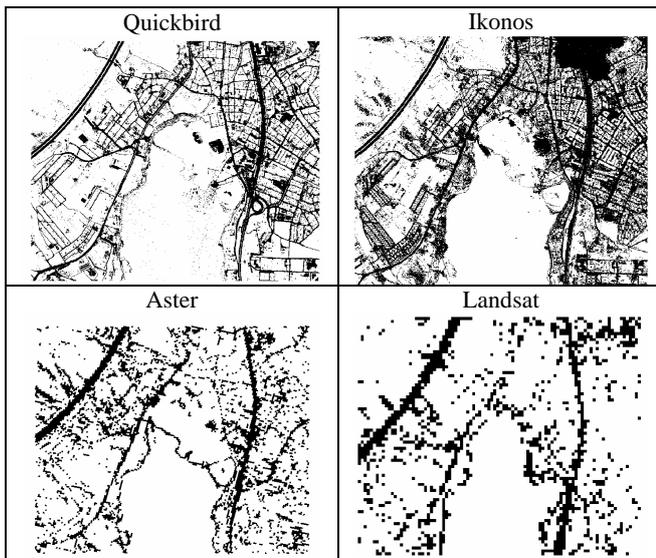


Figure 4: Classified images with two classes.

### 3. CONCLUSION

- Many objects which can be recognized from high resolution images can not be detected distinctly in a low resolution image.
- In a low resolution image roads can be extracted by using basic methods but the results have low accuracy.
- In a high resolution image many detail information about roads can be obtained and the roads can be detected more accurately. However complex methods are required for this process.

- Finally, the accuracy of road extraction from high resolution images is more than that one from low resolution images.

### REFERENCES

- Baltsavias, E., Gruen, A., van Gool, L. (Eds.), 2001. "Automatic Extraction of Man-Made Objects from Aerial and Space Images (III)". Balkema, Lisse, The Netherlands".
- Fortier, M.F., Ziou, D., Armenakis, C. and Wang, S., 1999. "Survey of Work on Road Extraction in Aerial and Satellite Images". Center for Topographic Information Geomatics, Ontario, Canada. Technical Report No. 241.
- Gruen, A., and H. Li, 1997. "Semiautomatic linear feature extraction by dynamic programming and LSB-snakes", *Photogrammetric Engineering & Remote Sensing*, 63(8):985-995.
- Gruen, A., Baltsavias, E., Henricsson, O. (Eds.), 1997. "Automatic Extraction of Man-Made Objects from Aerial and Space Images (II)". Birkhauser Verlag, Basel.
- Koike, K., Nagano, S. And Ohmi, M., 1995, "Lineament Analysis of Satellite Images Using A Segment Tracing Algorithm (STA)", *Computers and Geosciences*, Vol. 21, No. 9, 1091-1104.
- Ruisheng Wang ve Yun Zhang, 2003. "Extraction of urban road network using Quickbird Pansharped Multispectral and Panchromatic Imagery by Performing Edge-Aided Post-Classification" [http://www.geoict.net/Resources/Publications/RW\\_UNB\\_GRS\\_S\\_2003.pdf](http://www.geoict.net/Resources/Publications/RW_UNB_GRS_S_2003.pdf)
- Wiedemann, C., Hinz, S., 1999. "Automatic extraction and evaluation of road networks from satellite imagery". *International Archives of Photogrammetry and Remote Sensing*, vol. 32, Part 3-2W5, pp. 95-100.
- Xiangyun Hu C. Vincent TaoYong Hu, 2004 "Automatic road extraction from dense urban area by integrated processing of high resolution imagery and lidar data" <http://www.isprs.org/istanbul2004/comm3/papers/288.pdf>
- Zheng, Jeffrey, Liu, Gang-Jun, and Coppa, Isabel. 1998. "Linear Geo-spatial feature extraction from remotely sensed digital imagery using conjugate transformation". 9th Australian Remote Sensing and Photogrammetry Conference. vol. 1, pg 6401.

