Development of the Image Processing System Based on Hardware/Software Co-Design

王嘉宏、黃登淵

E-mail: 9606958@mail.dyu.edu.tw

ABSTRACT

Thresholding is widely used for image processing. Since each object has its own distinct gray-level distribution, an appropriate threshold is required to extract the objects from their background effectively. But the existing methods are quite complex, time-consuming, and hard to implement on the hardware efficiently. To overcome the inefficiency of the existing methods, this study proposes an algorithm called HGEM (Histogram-based Gaussian Estimation Method), and implements the Sobel edge detector on a Xilinx Virtex-4 (XC4VLX25) FPGA. Synthesized report shows that the proposed architecture on FPGA can run at a speed of 193.9 MHz, which is equivalent to a theoretical frame rate of 1,479 frames per second with gray-level images of 256 x 256 pixels. This result confirms that the proposed architecture on FPGA can easily achieve the requirements for a real-time image processing system. But for many different applications, bilevel threshold is often not enough for objects extracted from their background. Thus multilevel threshold selection is necessitated for such applications. The Otsu 's method can be easily extended from bilevel to multilevel threshold selection, but the computation complexity of the Otsu 's method exponentially increases with the increase in the class number of an image. To improve the inefficiency of the Otsu's method, this study proposes an algorithm called TSMO method (Two-Stage Multithreshold Otsu method), which is based on the Otsu 's method by maximizing the between-class variance. The TSMO method first divides the histogram into several groups with certain gray levels, and completes the number of cumulative pixels and the mean for each group statistically. Then the Otsu 's method can be applied to find the groups with the maximal between-class variances. Hence, the optimal thresholds can be found in those groups by applying Otsu's method again in a similar fashion. The results show that when the TSMO method with 32 group numbers (TSMO-32) is applied, the error relative to the Otsu 's method is less than 1%, but the performance of the TSMO-32 method can achieve more than 100,000 times compared to the Otsu's method with five thresholds.

Keywords: Edge detection; Otsu; HGEM; TSMO; FPGA

Table of Contents

封面內頁 簽名頁 授權書.		iii 中文摘要	iv 英文	【摘要	V
誌謝	vii 目錄	viii 圖	目錄	xi 表目	
録	xiv 第一章 緒論 1.1 研	究背景	11.2 研究目的	的	3 1.3 研
究方法	4 1.4 論文架構	6 第	第二章 影像邊緣偵測技	技術 2.1 邊緣偵測演	i算
法7 2.2	2 Otsu閥值演算法	11 2.3 R	ecursive Otsu演算法	12 2.4	HGEM演算
法14	12.5 TSMO演算法	19 第	三章 影像邊緣偵測閥個	值參數化之硬體架	構 3.1 系統架
構2	27 3.2 資料傳輸單元	28 3.3	BFPGA記憶體資源	29 3	3.3.1 FIFO記憶
體電路設計	30 3.4 影像資料提取單	.元3	85 3.4.1 資料提取單元]	功能模擬	38 3.4.2 資
料提取單元資源使用分析	T 40 3.5 Sobel:	運算單元	44 3.6 閥值運算	章單元	45 第四
章 Xilinx MicroBlaze系統	開發介紹 4.1 MicroBlaze	處理器	48 4.2 晶片匯流排	簡介	50 4.2.1
CoreConnect系統架構	51 4.2.2 LME	8匯流排	53 4.3 MicroBla	aze系統架構	55 4.4
OPB匯流排之矽智財連接	接介面 56 4.5	系統雛形建立	58 第五章	實驗結果與討論	5.1 HGEM演
算法分析與討論	61 5.2 TSMO演算	法分析與討論	63 5.3 硬體架	構分析	71 第
六章 結論與未來發展方向	句 6.1 結論	75 6.2 未芽	マ研究方向	76 參考文	
獻	78				

REFERENCES

- [1] K. Benkrid, D. Crookes, A. Benkrid, Towards a general framework for FPGA based image processing using hardware skeletons, Journal of Parallel Computing 28 (7-8) (2002) 1141-1154.
- [2] Xue Li, Rongchun Zhao, Qing Wang, FPGA based Sobel algorithm as vehicle edge detector in VCAS, IEEE International Conference on Neural Networks and Signal Processing, Nanjing, China, December (2003) 1139-1142.

- [3] R.L. Rosas, A. de Luca, F.B. Santillan, SIMD architecture for image segmentation using Sobel operators implemented in FPGA technology, The 2nd International Conference on Electrical and Electronics Engineering (ICEEE) and XI Conference on Electrical Engineering, Mexico City, Mexico, September (2005) 77-80.
- [4] P.Y. Hsiao, C.H. Chen, H. Wen, S.J. Chen, Real-time realisation of noise-immune gradient-based edge detector, IEE Proceedings-Computers and Digital Techniques, 153 (4) (2006) 261-269.
- [5] A. Benedetti, A. Prati, N. Scarabottolo, Image convolution on FPGAs: the implementation of a multi-FPGA FIFO structure, Proceedings of the 24th Euromicro Conference, Vasteras, Sweden, August (1998) 123-130.
- [6] B. Bosi, G. Bois, Y. Savaria, Reconfigurable pipelined 2-D convolvers for fast digital signal processing, IEEE Transactions on Very Large Scale Integration (VLSI) Systems 7 (3) (1999) 299-308.
- [7] V. Muthukumar, D.V. Rao, Image processing algorithms on reconfigurable architecture using HandelC, Proceedings of the Euromicro Symposium on Digital System Design, Rennes, France (2004) 218-226.
- [8] C.K. Lee, F.W. Choy, H.C. Lam, Real-time thresholding using histogram concavity, Proceedings of the IEEE International Symposium on Industrial Electronics, Xian, China, 1 (1992) 500-503.
- [9] S.E. El-Khamy, M. Lotfy, N. El-Yamany, A modified Fuzzy Sobel edge detector, Seventeenth National Radio Science Conference, Minufiya University, Egypt, February (2000) C32 1-9.
- [10] Jianping Fan, Walid G. Aref, Mohand-Said Hacid, Ahmed K. Elmagarmid, An improved automatic isotropic color edge detection technique, Journal of Pattern Recognition Letters 22 (13) (2001) 1419-1429.
- [11] C.H. Li, P.K.S. Tam, An iterative algorithm for minimum cross-entropy thresholding, Pattern Recognition Letter 19 (1998) 771-776.
- [12] Shu Yang, Ying Han, Cai-Rong Wang, Xiao-Wei Wang, Fast selecting threshold algorithm based on one-dimensional entropy, Proceedings of the Fourth International Conference on Machine Learning and Cybernetics, Guangzhou, China, August (2005) 4554-4557.
- [13] URL: http://www.csse.monash.edu.au/hons/projects/2002/Laura.Frost/index.html [14] R.C. Gonzalez, R.E. Woods, Digital image processing, 2nd edition, Prentice Hall, Upper Saddle River, New Jersey, 2002.
- [15] A.T. Abak, U. Baris, B. Sankur, The performance evaluation of thresholding algorithms for optical character recognition, Proceedings of the Fourth International Conference on Document Analysis and Recognition August (1997), 697-700.
- [16] Feixiang Yan, Hong Zhang, C. Ronald Kube, A multistage adaptive thresholding method, Pattern Recognition Letters 26 (8) (2005), 1183-1191.
- [17] D. Aiteanu, D. Ristic, A. Graser, Content based threshold adaptation for image processing in industrial application, International Conference on Control and Automation (ICCA2005), June (2005), 1022-1027.
- [18] Hui-Fuang Ng, Automatic thresholding for defect detection, Pattern Recognition Letters 27 (14) (2006), 1644-1649.
- [19] E.P. Ong, B.J. Tye, W.S. Lin, M. Etoh, An efficient video object segmentation scheme, Proceedings of IEEE International Conference on Acoustics, Speech, and Signal Processing (ICASSP2002), (2002), IV-3361-IV-3364.
- [20] Guo Jing, D. Rajan, Chng Eng Siong, Motion Detection with Adaptive Background and Dynamic Thresholds, 2005 Fifth International Conference on Information. Communications and Signal Processing. December (2005).41-45.
- [21] C. Su, A. Amer, A Real-Time Adaptive Thresholding for Video Change Detection, 2006 IEEE International Conference on Image Processing, October (2006), 157-160.
- [22] O. Sukmarg, K.R. Rao, Fast object detection and segmentation in MPEG compressed domain, Proceedings of TENCON 2000 (2000), 364-368.
- [23] D. Zhang, G. Lu, Segmentation of Moving Objects in Image Sequence: A Review, Journal of Circuits, Systems, and Signal Processing 20 (2) (2001), 143-183.
- [24] A. Amer, Memory-based spatio-temporal real-time object segmentation, Proc. SPIE Int. Symposium on Electronic Imaging, Conf. on Real-Time Imaging (RTI), April (2003), 10-21.
- [25] Shao-Yi Chien, Yu-Wen Huang, Bing-Yu Hsieh, Shyh-Yih Ma, Liang-Gee Chen, Fast video segmentation algorithm with shadow cancellation, global motion compensation, and adaptive threshold techniques, IEEE Transactions on Multimedia 6 (5) (2004), 732-748.
- [26] M.S. Atkins, B.T. Mackiewich, Fully Automatic Segmentation of the Brain in MRI, IEEE Transactions on Medical Imaging 17 (1) (1998), 98-107.
- [27] P.K. Saha, J.K. Udupa, Optimum Image Thresholding via Class Uncertainty and Region Homogeneity, IEEE Transactions on Pattern Analysis and Machine Intelligence 23 (7) (2001), 689-706.
- [28] N. Otsu, A threshold selection method from gray-level histogram, IEEE Transactions on System, Man, and Cybernetics, SMC-9 (1) (1979) 62-66.
- [29] P.S. Liao, T.S. Chen, P.C. Chung, A fast algorithm for multilevel thresholding, Journal of Information Science and Engineering 17 (5) (2001), 713 727.
- [30] B. Sankur, M. Sezgin, Survey over image thresholding techniques and quantitative performance evaluation, Journal of Electronic Imaging 13 (1) (2004) 146-165.

[31] W.A. Yasnoff, J.K. Mui, J.W. Bacus, Error measures for scene segmentation, Pattern Recognition 9(1997), 217 – 231.
[32] CEDAR Database of Handwritten words, URL: http://www.cedar.buffalo.edu/Databases/CDROM1/words.html [33] FVC2000 Fingerprint Database, URL: http://bias.csr.unibo.it/fvc2000/db2.asp