























a tensor of size  $N \times N \times N$  is represented as an outer product of 3 mutually orthogonal  $N$ -dimensional vectors, while for the 3DHSVD the component is defined by the kernels of size  $2 \times 2 \times 2$ , contained by it. The basic advantages of the new decomposition for multi-dimensional images are the low computational complexity and the tree-like structure, which permits cutting-off the low-energy branches through threshold selection. This makes the algorithm suitable for parallel processing with multi-processor systems, where the basic processor executes the kernel decomposition algorithm for a tensor of size  $2 \times 2 \times 2$ . The 3DHSVD algorithm opens new possibilities for fast image processing in various application areas, such as: image filtration, segmentation, merging, digital watermarking, extraction reduced number of features for pattern recognition, etc.

## References

- [1] Fieguth, P. (2011). *Statistical image processing and multidimensional modeling*, Springer.
- [2] Diamantaras, K., Kung, S. (1996). *Principal Component Neural Networks*, Wiley, New York.
- [3] Orfanidis, S. (2007). *SVD, PCA, KLT, CCA, and All That*, Rutgers University Electrical & Computer Engineering Department, Optimum Signal Processing, pp. 1-77.
- [4] Levy, A., Lindenbaum, M. (2000). *Sequential Karhunen-Loeve Basis Extraction and its Application to Images*, IEEE Trans. on Image Processing, Vol. 9, No. 8, pp. 1371-1374.
- [5] Sadek, R. (2012). *SVD Based Image Processing Applications: State of The Art*, Intern. J. of Advanced Computer Science and Applications, 3(7), pp. 26-34.
- [6] Drinea, E., Drineas P., Huggins, P. (2001). *A Randomized SVD Algorithm for Image Processing Appl.*, Proc. of the 8<sup>th</sup> Panhellenic Conf. on Informatics, Y. Manolopoulos, S. Evripidou (Eds), Nicosia, Cyprus, pp. 278-288.
- [7] Holmes, M., Gray, A., Isbell, C. (2008). *QUIC-SVD: Fast SVD using Cosine trees*, Proc. of NIPS, pp. 673-680.
- [8] Foster, B., Mahadevan, S., Wang, R. (2012). *A GPU-based Approximate SVD Algorithm*, 9<sup>th</sup> Intern. Conf. on Parallel Processing and Applied Mathematics, 2011, LNCS, Vol. 7203, pp. 569-578.
- [9] Yoshikawa, M., Gong, Y., Ashino, R., Vaillancourt, R. (2005). *Case study on SVD multiresolution analysis*, CRM-3179, pp. 1-18.
- [10] Waldemar, P., Ramstad, T. (1997). *Hybrid KLT-SVD image compression*, IEEE Intern. Conf. on Acoustics, Speech, and Signal Processing, IEEE Comput. Soc. Press, Los Alamitos, pp. 2713-2716.
- [11] Aharon, M., Elad, M., Bruckstein, A. (2006). *The K-SVD: an algorithm for designing of overcomplete dictionaries for sparse representation*, IEEE Trans. on Signal Processing, 54, pp. 4311-4322.
- [12] Singh, S., Kumar, S. (2011). *SVD Based Sub-band Decomposition and Multiresolution Representation of Digital Colour Images*, Pertanika J. of Science & Technology, 19 (2), pp. 229-235.
- [13] Kountchev, R., Kountcheva, R. (2015). *Hierarchical SVD for Halftone Images*. The 7<sup>th</sup> International Conference on Information Technology Big Data, Al Zaytoonah University, Amman, Jordan, May 12-15, pp. 50-58.
- [14] Kountchev, R., Kountcheva, R. (2015). *Hierarchical SVD-based Image Decomposition with Tree Structure*, Intern. J. of Reasoning-Based Intelligent Systems, 7(1/2), pp.114-129.
- [15] De Lathauwer, L. (2008). *Decompositions of a higher-order tensor in block terms - Part I and II*, SIAM J. Matrix Anal. Appl., Vol. 30, pp. 1022-1066.
- [16] Kolda, T., Bader, B. (2009). *Tensor decompositions and applications*, SIAM Review, 51(3), pp. 455-500.
- [17] Bergqvist, G., Larsson, E. (2010). *The Higher-Order SVD: Theory and an Application*, IEEE Signal Processing Magazine, 27(3), pp. 151-154.
- [18] Salmi, J., Richter, A., Koivunen, V. (2009). *Sequential Unfolding SVD for Tensors with Applications in Array Signal Processing*, IEEE Trans, on Signal Processing, 57 (12), pp. 4719-4733.
- [19] Grasedyck, L. (2010). *Hierarchical SVD of Tensors*, SIAM J, on Matrix Analysis and Applications, 31(4), pp. 2029-2054.
- [20] Oseledets, I. (2011). *Tensor-train decomposition*, SIAM Journal on Scientific Computing, 33(5), pp. 2295-2317.
- [21] Cichocki, A., Mandic, D., Phan, A-H., Caiafa, C., G. Zhou, Zhao, Q., De Lathauwer, L. (2015). *Tensor Decompositions for Signal Processing Applications*, IEEE Signal Processing Magazine, 32(2), pp. 145-163.
- [22] Wu, Q., Xia, T., Yu, Y. (2007). *Hierarchical Tensor Approximation of Multidimensional Images*, IEEE Intern. Conf. on Image Processing (ICIP'07), San Antonio, TX, Vol. 6, pp. IV-49, IV-52.