Author's Accepted Manuscript

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www.elsevier.com/locate/pr

PII: S0031-3203(16)30324-7

DOI: http://dx.doi.org/10.1016/j.patcog.2016.10.009

Reference: PR5917

To appear in: Pattern Recognition

Received date: 24 May 2016

Revised date: 13 September 2016 Accepted date: 6 October 2016

Cite this article as: Tong Tong, Katherine Gray, Qinquan Gao, Liang Chen and Daniel Rueckert, Multi-Modal Classification of Alzheimer's Disease Using Nonlinear Graph Fusion, *Pattern Recognition* http://dx.doi.org/10.1016/j.patcog.2016.10.009

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ACCEPTED MANUSCRIPT

Multi-Modal Classification of Alzheimer's Disease Using Nonlinear Graph Fusion

Tong Tong^{a,c,}, Katherine Gray^a, Qinquan Gao^b, Liang Chen^a, Daniel Rueckert^a, and the Alzheimer's Disease Neuroimaging Initiative¹

Abstract

Accurate diagnosis of Alzheimer's disease (AD) and its prodromal stage mild cognitive impairment (MCI) is of great interest to patients and clinicians. Recent studies have demonstrated that multiple neuroimaging and biological measures contain complementary information for diagnosis and prognosis. Classification methods are needed to combine these multiple biomarkers to provide an accurate diagnosis. State-of-the-art approaches calculate a mixed kernel or a similarity matrix by linearly combining kernels or similarities from multiple modalities. However, the complementary information from multi-modal data are not necessarily linearly related. In addition, this linear combination is also sensitive to the weights assigned to each modality. In this paper, we present a multi-modality classification framework to efficiently exploit the complementarity in the multi-modal data. First, pairwise similarity is calculated for each modality individually using the features including regional MRI volumes, voxel-based FDG-PET signal intensities, CSF biomarker measures, and categorical genetic information. Similarities from multiple modalities are then combined

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¹Data used in the preparation of this article were obtained from the ADNI database (www.loni.ucla.edu/ADNI). As such, the investigators within the ADNI contributed to the design and implementation of ADNI and/or provided data but did not participate in analysis or writing of this report. A complete listing of ADNI investigators can be found at: www.loni.ucla.edu/ADNI/Collaboration/ADNI_Authorship_list.pdf

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