
Mixed Importance Weighting for Domain Adaptation under Covariate Shift

David S. Hippocampus*
Department of Computer Science
Cranberry-Lemon University
Pittsburgh, PA 15213
hippo@cs.cranberry-lemon.edu

Abstract

Pass.

1 Introduction

Citations of inproceedings [1].

Citations of book [2].

Citations of article [3].

References

- [1] Alexander , J. & Mozer , M. C. (1995) Template-based algorithms for connectionist rule extraction. In G. Tesauro, D. S. Touretzky, and T. K. Leen, (eds.), *Advances in Neural Information Processing Systems 7*, pp. 609–616, Cambridge, MA: MIT Press.
- [2] Bower , J. M. & Beeman , D. (1995) *The Book of GENESIS: Exploring Realistic Neural Models with the GEneral NEural Simulation System*. New York: TELOS/Springer–Verlag.
- [3] Hasselmo , M. E., Schnell , E., & Barkai , E. (1995) Dynamics of learning and recall at excitatory recurrent synapses and cholinergic modulation in rat hippocampal region ca3. *Journal of Neuroscience* **15**(7):5249–5262.

References

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- [1] Alexander, J.A. & Mozer, M.C. (1995) Template-based algorithms for connectionist rule extraction. In G. Tesauro, D.S. Touretzky and T.K. Leen (eds.), *Advances in Neural Information Processing Systems 7*, pp. 609–616. Cambridge, MA: MIT Press.
- [2] Bower, J.M. & Beeman, D. (1995) *The Book of GENESIS: Exploring Realistic Neural Models with the GEneral NEural Simulation System*. New York: TELOS/Springer–Verlag.
- [3] Hasselmo, M.E., Schnell, E. & Barkai, E. (1995) Dynamics of learning and recall at excitatory recurrent synapses and cholinergic modulation in rat hippocampal region CA3. *Journal of Neuroscience* **15**(7):5249–5262.

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