Introduction,
Well-Posed Learning Problems,
& Concept Learning

Relevant Readings: Sections 1.1, 2.1-2.2 in Mitchell

CS495 - Machine Learning, Fall 2009

Introductions

- Pass out syllabus

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