Text Classification with RTextTools Odum Institute University of North Carolina

Loren Collingwood¹ (with Tim Jurka, Amber Boydstun, Emiliano Grossman, and Wouter van Atteveldt) February 6, 2012

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- Based on scientific method; establishes objectivity via inter-coder reliability

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What is Machine Learning?

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What the Data May Look Like

Article_ID	Date	Title	Subject	Topic Code
41246	1-Jan-96	Nation's Small	Jails overwhel	12
41257	2-Jan-96	FEDERAL IMPA	Federal budge	20
41268	3-Jan-96	Long, Costly P	Contenders for	20
41279	4-Jan-96	Top Leader of	Bosnian Serb	19

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Thank You

• Any questions contact Loren Collingwood lorenc2@uw.edu