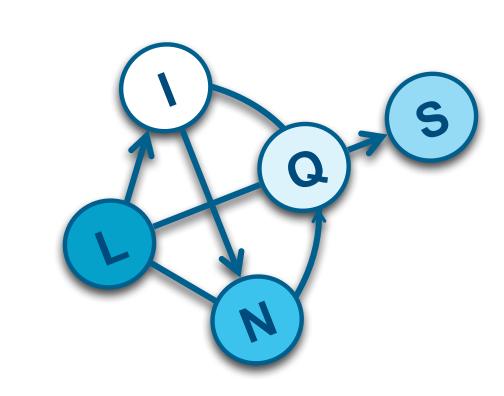


# Learning Latent Groups with Hinge-loss Markov Random Fields

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#### Introduction

- Can we build **tractable** models that learn latent mixed-memberships in rich data?
- We achieve this goal using a graphical model that encodes complex, yet interpretable dependencies among
  - group membership
  - language usage
  - social interactions
- We demonstrate our approach by learning indicators of latent political preferences

## Hinge-loss Markov Random Fields

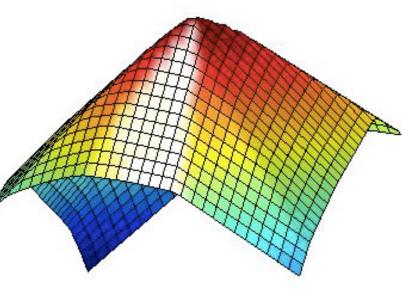
Hinge-loss Markov random fields are undirected graphical models analogous to discrete MRFs.

- Variables are continuous valued in [0,1]
- Potentials are hinge-loss functions

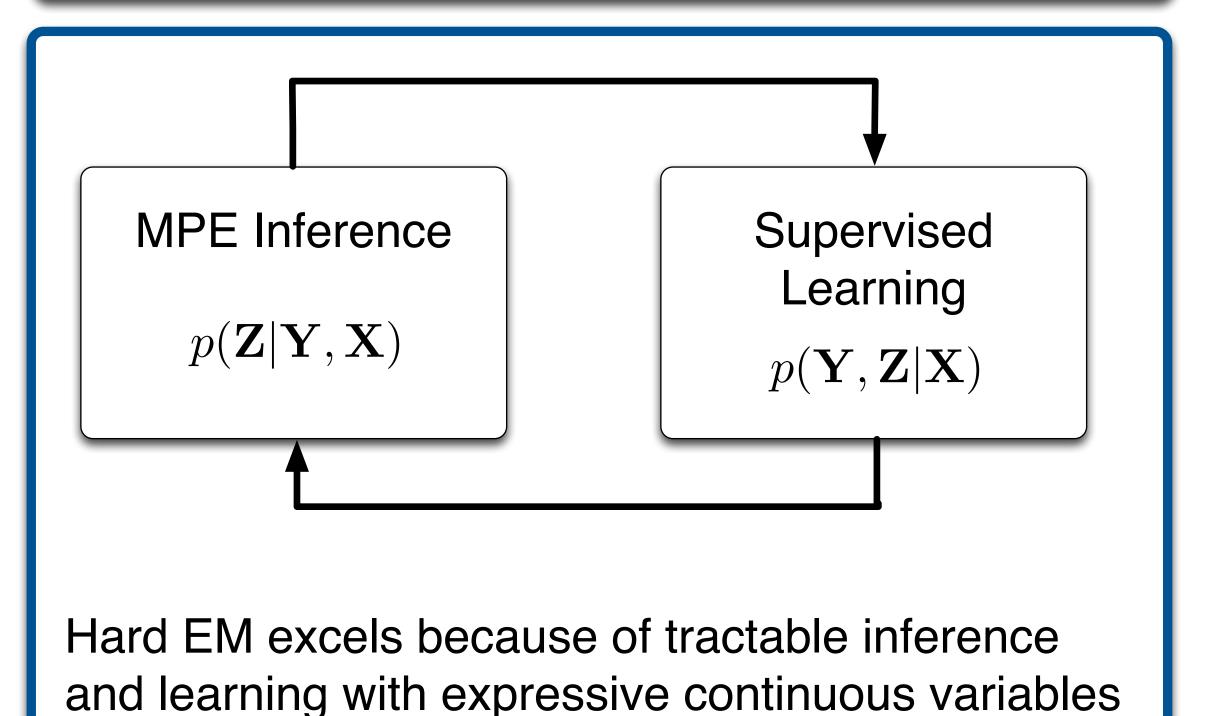
$$p(\mathbf{Y}|\mathbf{X}) = \frac{1}{Z} \exp \left[ -\sum_{j=1}^{m} w_j \max \{\ell_j(\mathbf{Y}, \mathbf{X}), 0\}^{p_j} \right]$$

#### <u>Features</u>

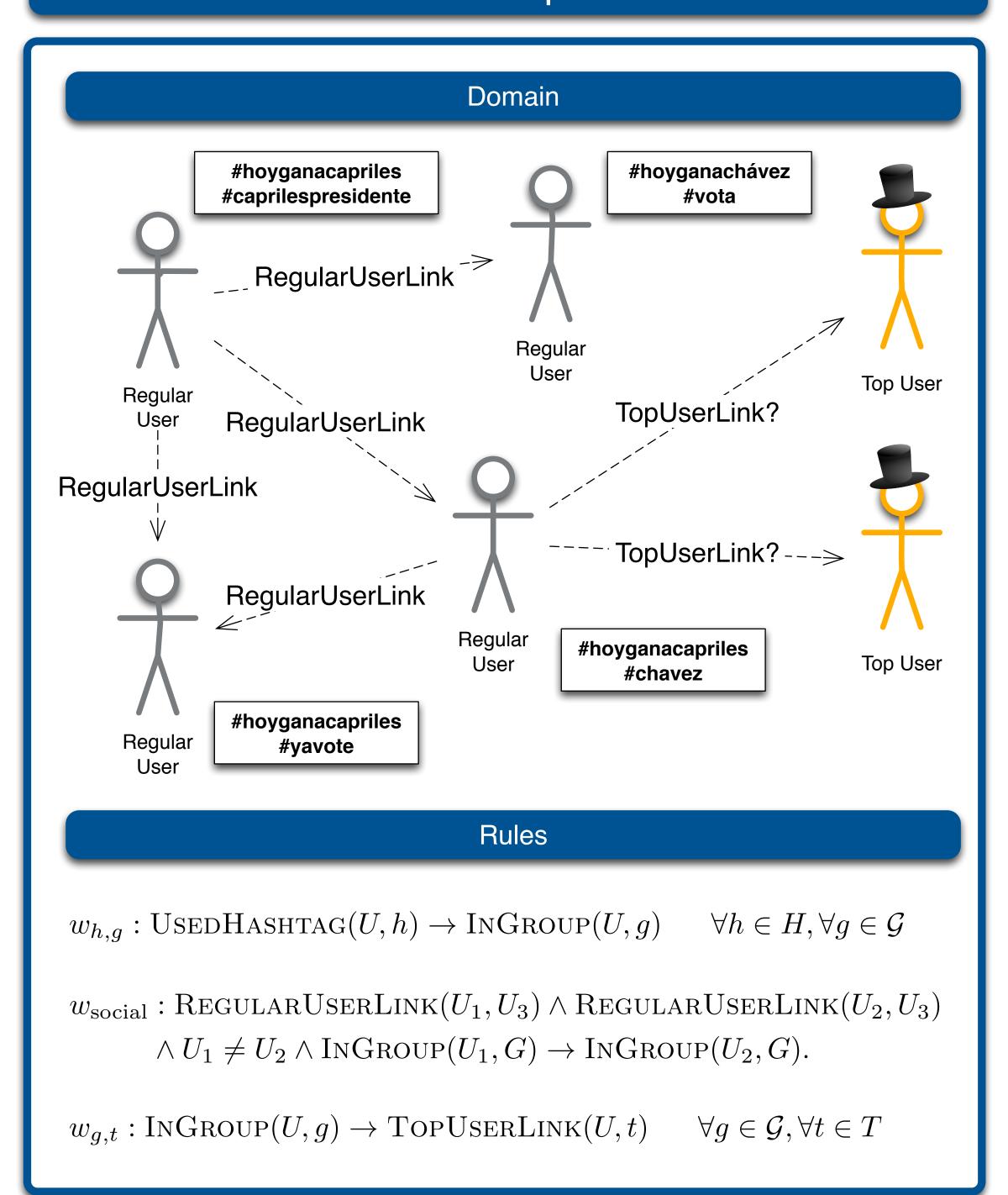
- Easily constructed and interpreted by mapping between logical rules and hingeloss functions
- Linear constraints enable mixture models
- Log-concave density function admits fast,
  exact MPE inference
  using the alternating
  direction method of
  multipliers (ADMM)



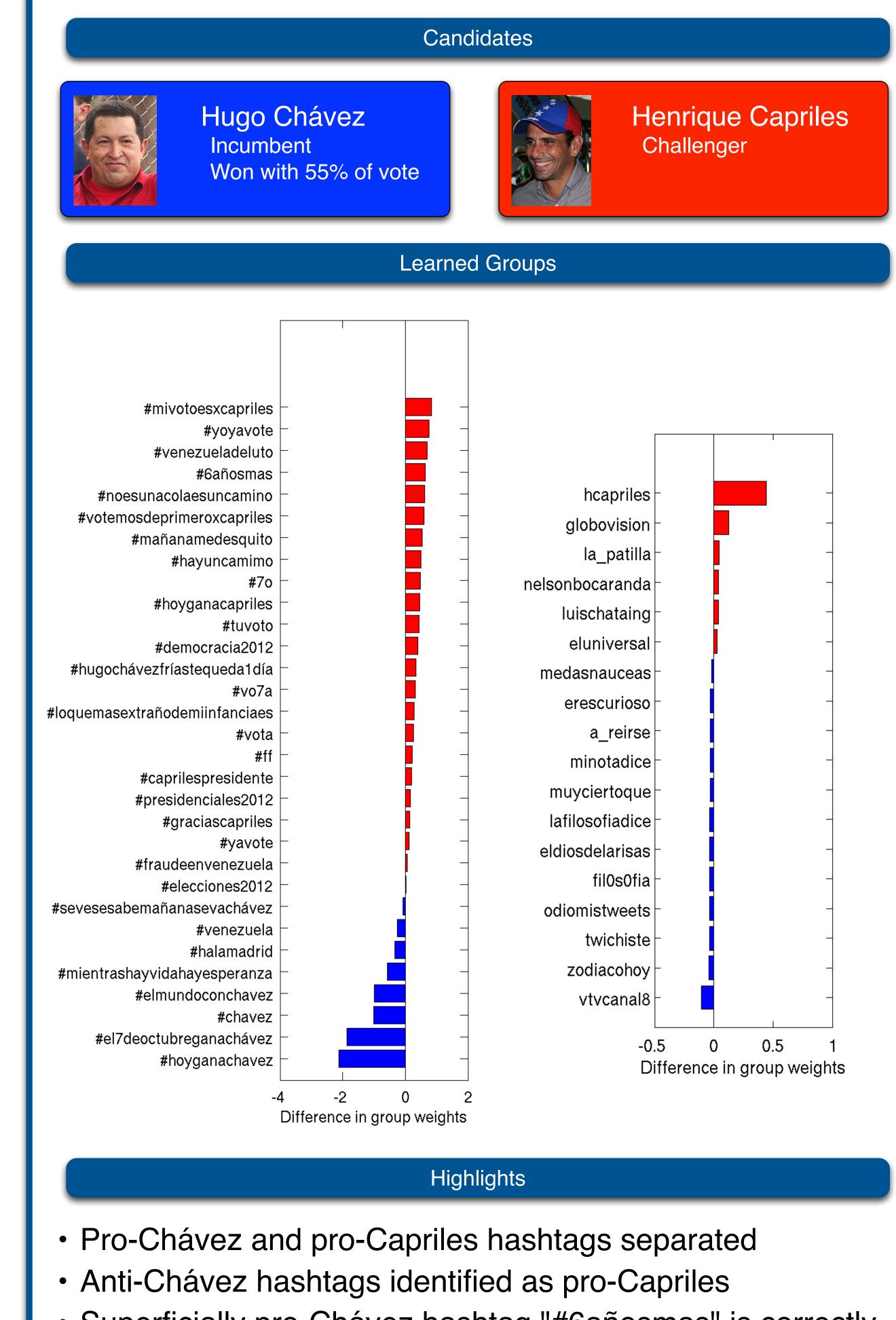
## Latent-Variable Learning



### Latent-Group Model



## Case Study: 2012 Venezuelan Presidential Election



- Superficially pro-Chávez hashtag "#6añosmas" is correctly identified as pro-Capriles
- Capriles's account associated with pro-Capriles group
- Independent and state-owned media accounts separated