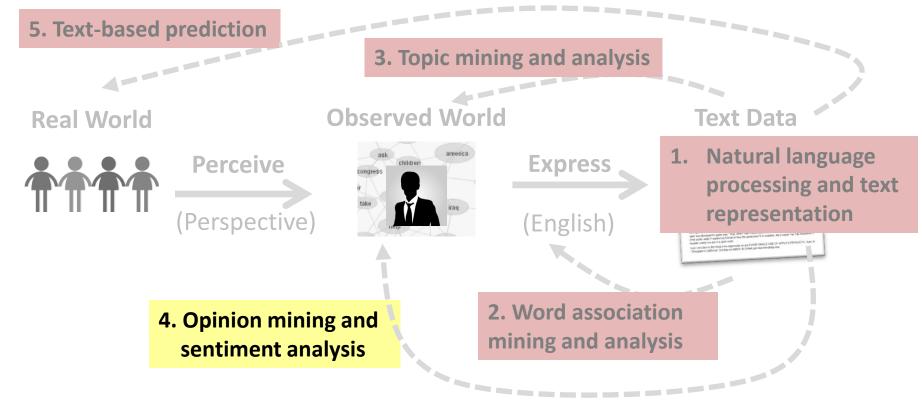
Opinion Mining and Sentiment Analysis: Latent Aspect Rating Analysis

Part 1

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Motivation

Hotel XYX

How to infer aspect ratings?

Reviewer 1: ★★★★★

"Great location + spacious room = happy traveler"
Stayed for a weekend in July. Walked everywhere,
enjoyed the comfy bed and quiet hallways....



Reviewer 2: ★★★★ "Terrific service and gorgeous facility"

I stayed at the hotel wiht my young daughter for three nights June 17-20, 2010 and absolutely loved the hotel. The room was one of the nicest I've ever stayed in ...





Latent Aspect Rating Analysis [Wang et al. 10]

- Given a set of review articles about a topic with overall ratings
- Output
 - Major aspects commented on in the reviews
 - Ratings on each aspect
 - Relative weights placed on different aspects by reviewers
- Many applications
 - Opinion-based entity ranking
 - Aspect-level opinion summarization
 - Reviewer preference analysis
 - Personalized recommendation of products

— ...

Solving LARA in Two Stages 2.36+3.59 Latent Rating Regression Aspect Segmentation rd $\alpha_i(d)$ $c_i(wd)$ $\beta_{i,w}$ $r_i(d)$ **** Term Weights | Aspect Rating | Aspect Weight Aspect segments "A friend and I stayed at the Hotel \dots 0.0 location:1 The hotel was very nice. The location amazing:1 3.9 3.8 0.2 was amazing. We could walk almost 0.1 walk:1 anywhere, but ... far. The room was -0.2 far:1 very nicely appointed and the bed 0.1 room:1 1.7 was sooo comfortable. Even though nicely:1 0.2 4.8 0.1 the bathroom door did not close all appointed:1 3.9 the way, it was still pretty private. ... comfortable:1 But what I liked best about the hotel nice:1 2.1 was the staff. They were soooo nice accommod.:1 1.2 0.6 smile:1 1.7 and accommodating ..." friendliness:1 1.2 **Observed** attentiveness:1 Latent!

Latent Rating Regression [Wang et al. 10]

- Data: a set of review documents with overall ratings: C={(d, r_d)}

 d is pre-segmented into k aspect segments
 c_i(w_j d) = count of word w in aspect segment i (zero if w didn't occur)

 Model: predict rating based on d: p(r_d | d)

 Multivariate

 $r_{d} \sim N(\sum_{i=1}^{k} \alpha_{i}(d)r_{i}(d),\underline{\delta^{2}}), \qquad \qquad \vec{\alpha}(d) \sim N(\vec{\mu},\Sigma)$ $r_{i}(d) = \sum_{w \in V} c_{i}(w,d)\underline{\beta_{i,w}} \qquad \beta_{i,w} \in \Re$ Aspect-Specific Sentiment of w

Aspect Rating = Sum of sentiment weights of words in the aspect

Latent Rating Regression (cont.)

Maximum Likelihood Estimate

– Parameters: $\Lambda = (\{\beta_{i,w}\}, \bar{\mu}, \Sigma, \delta^2)$

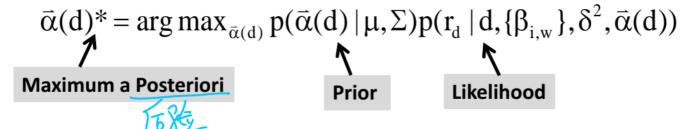
- ML estimate: $\Lambda^* = arg \max_{\Lambda} \prod_{d \in C} p(r_d \mid d, \Lambda)$

Aspect Rating for aspect i

$$r_i(d) = \sum_{w \in V} c_i(w, d)\beta_{i,w}$$

c_i(w,d)=0 for words not occurring in aspect segment i

Aspect Weights: α_i(d) = weight on aspect i



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Suggested Reading

 [Wang et al. 10] Hongning Wang, Yue Lu, and ChengXiang Zhai, Latent aspect rating analysis on review text data: a rating regression approach. In Proceedings of ACM KDD 2010, pp. 783-792, 2010. DOI=10.1145/1835804.1835903