

# RAND-WALK: A latent variable model approach to word embeddings

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Presented by  
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# Pointwise Mutual Information (PMI)

Church & Hanks 1990

$$\text{PMI}(w, w') = \log \frac{p(w, w')}{p(w)p(w')}$$

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Empirically low rank!

$$\langle v_w, v_{w'} \rangle \approx \text{PMI}(w, w')$$

# This Paper

- Science: Why do we have low rank?

What is the sensible data generating process leading to this approximate low rank behavior?

# Word Embeddings as Metric Recovery in Semantic Spaces

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$$P(X_t = j | X_{t-1} = i) = h\left(\frac{1}{\sigma} \|x_i - x_j\|_2^2\right)$$

Random walk on the word embedding directly

# Log-linear word production model

- This paper assumes discourse/context  $c_t$

$$\Pr[w \text{ emitted at time } t \mid c_t] \propto \exp(\langle c_t, v_w \rangle)$$

# Discourse Random Walk

- Slow random walk
- Stationary being Gaussian

Isotropy

$$v = s \cdot \hat{v}$$

# Recipe

Log-linear selection



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Isotropy of  $v$

Gaussian Stationary of  $c$



$$\Pr[w \text{ emitted at time } t \mid c_t] = \frac{\exp(\langle c_t, v_w \rangle)}{\text{constant}}$$

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$$\begin{aligned} p(w, w') &= \mathbb{E}_{c, c'} [\Pr[w, w' \mid c, c']] \\ &= \mathbb{E}_{c, c'} [p(w \mid c)p(w' \mid c')] \end{aligned}$$

# Recipe

Log-linear selection   $\Pr[w \text{ emitted at time } t \mid c_t] \propto \exp(\langle c_t, v_w \rangle)$

Isotropy of  $v$   
Gaussian Stationary of  $c$    $\Pr[w \text{ emitted at time } t \mid c_t] = \frac{\exp(\langle c_t, v_w \rangle)}{\text{constant}}$

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Small Walk of  $c$    $\mathbb{E}[\exp(\langle v_w + v_{w'}, c \rangle)].$

# Recipe

Log-linear selection  $\longrightarrow$   $\Pr[w \text{ emitted at time } t \mid c_t] \propto \exp(\langle c_t, v_w \rangle)$

Isotropy of  $v$   
Gaussian Stationary of  $c$   $\longrightarrow$   $\Pr[w \text{ emitted at time } t \mid c_t] = \frac{\exp(\langle c_t, v_w \rangle)}{\text{constant}}$

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Small Walk of  $c$   $\longrightarrow$   $\mathbb{E}[\exp(\langle v_w + v_{w'}, c \rangle)]$ .

Gaussian Stationary of  $c$   $\longrightarrow$   $\exp\left(\frac{\|v_w + v_{w'}\|^2}{2d}\right)$

# Relation

Practice

*“man:woman::king:??,”*

$$v_{king} - v_{man} \dagger v_{woman}.$$

The paper **probably** explains that we want the change of measure density ratio for the two changes to be similar

# Discussion

We assume

- 1. Log-linear selection
- 2. Small discourse walk
- 3. Isotropy

For relation, we assume the goal is somewhat linear(?)