

# Taking uncertainty seriously

A Bayesian approach to word embedding bias estimation

Alicja Dobrzeniecka & Rafal Urbaniak  
(LoPSE research group, University of Gdansk)

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# Word2vec

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Well, you could use 30k binary vectors with a slot for each lexical unit. . .  
. . . . . but this would be inefficient and wouldn't capture any relations  
between words.

# Word2vec

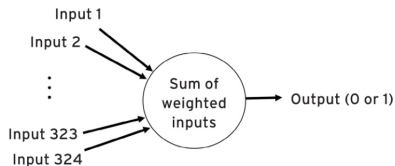
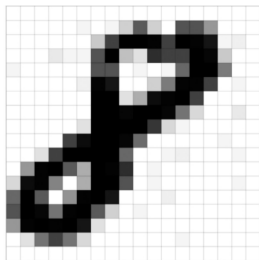


Illustration: M. Mitchell

## Rosenblatt's perceptron

- Inputs (pixel intensities) with weights
- Nodes with activation levels from 0-1
- (Perhaps) 0-1 output based on a threshold

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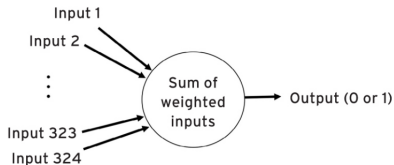
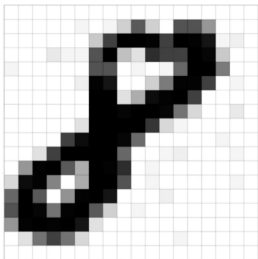


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## Learning

- Start with random weights
- Test on a case:
  - If right, don't change weights.
  - If wrong, change weights a bit, with focus on the ones more responsible for the judgment:

$$w_j \leftarrow w_j = \underbrace{\eta}_{\text{learning rate}} \left( \underbrace{t}_{\text{correct output}} - \underbrace{y}_{\text{actual output}} \right) \underbrace{x_j}_{\text{actual input}}$$

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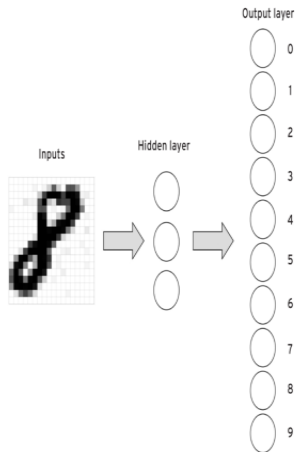


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- Each hidden unit takes a weighted sum of 324 inputs and passes on its activation level as input to outer layer units.
- Activation levels of outer layers are interpreted as network's levels of confidence in a classification problem.
- Learning: back-propagation (gradient descent: approximate the direction of steepest descent in the error surface w.r.t to weights, modify accordingly).

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## Distributional semantics

- "You shall know a word by the company it keeps" (John Firth, 1957)
- "the degree of semantic similarity between two linguistic expressions  $A$  and  $B$  is a function of the similarity of the linguistic contexts in which  $A$  and  $B$  can appear." (A. Lenci, 2008)



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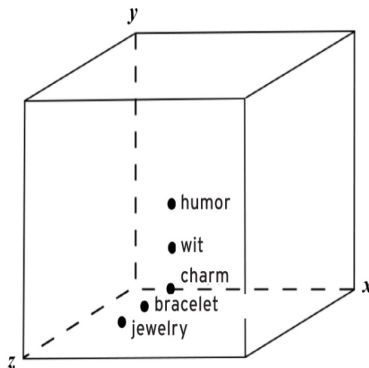


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## Google and Mikolov

*Efficient Estimation of Word Representation in Vector Space*, 2013

Let's train a neural network and use vectors of weights!

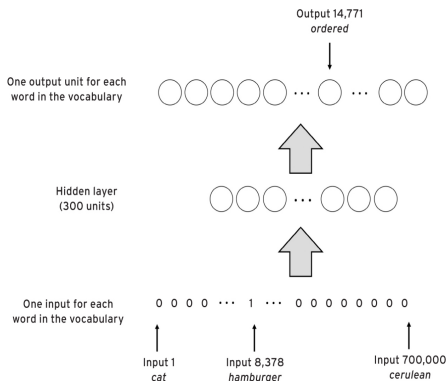


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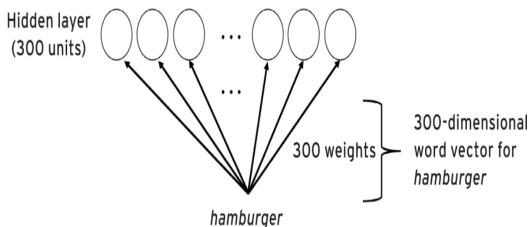


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## Nearest words

- **philosophy**: philosophies, credo, ethos, principles, ethic, tenets, mantra, ideology, mindset, worldview
- **sandwich**: sandwiches, burger, chicken sandwich, cheeseburger, burrito, burgers, pizza, turkey sandwich, hamburger, burritos

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## Some similarities from *philosophy*

Logic (.47), Nietzsche (.32), Hegel (.32), analytic (.13), burger (.08), continental (.04), Russell (.04)