

COMP90042

Web search and text analysis

Workshop Week 6

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https://github.com/HanXudong/COMP90042_Workshops

Review

1. N-gram model
2. Backoff and interpolation

1. <s> <s> how much wood would a wood chuck chuck if
a wood chuck would chuck wood </s>
2. <s> <s> a wood chuck would chuck the wood he could
chuck if a wood chuck would chuck wood </s>

<s> <s> a wood chuck would chuck the wood he could chuck if a
wood chuck would chuck wood </s></s>

$$P(w_1, w_2, \dots, w_m) = \prod_{i=1}^m P(w_i | w_{i-2} w_{i-1})$$

$$P_{add1}(w_i | w_{i-2} w_{i-1}) = \frac{C(w_{i-2} w_{i-1} w_i) + 1}{C(w_{i-2} w_{i-1}) + V}$$

Q3: What does back-off mean, in the context of smoothing a language model? What does interpolation refer to?

- The idea in a Backoff model is to build an Ngram model based on an (N-1) model
- https://en.wikipedia.org/wiki/Katz%27s_back-off_model
- Interpolation: instead of just backing off to the non-zero Ngram, it is possible to take into account all Ngrams.
- Estimate lambdas from held-out dataset.

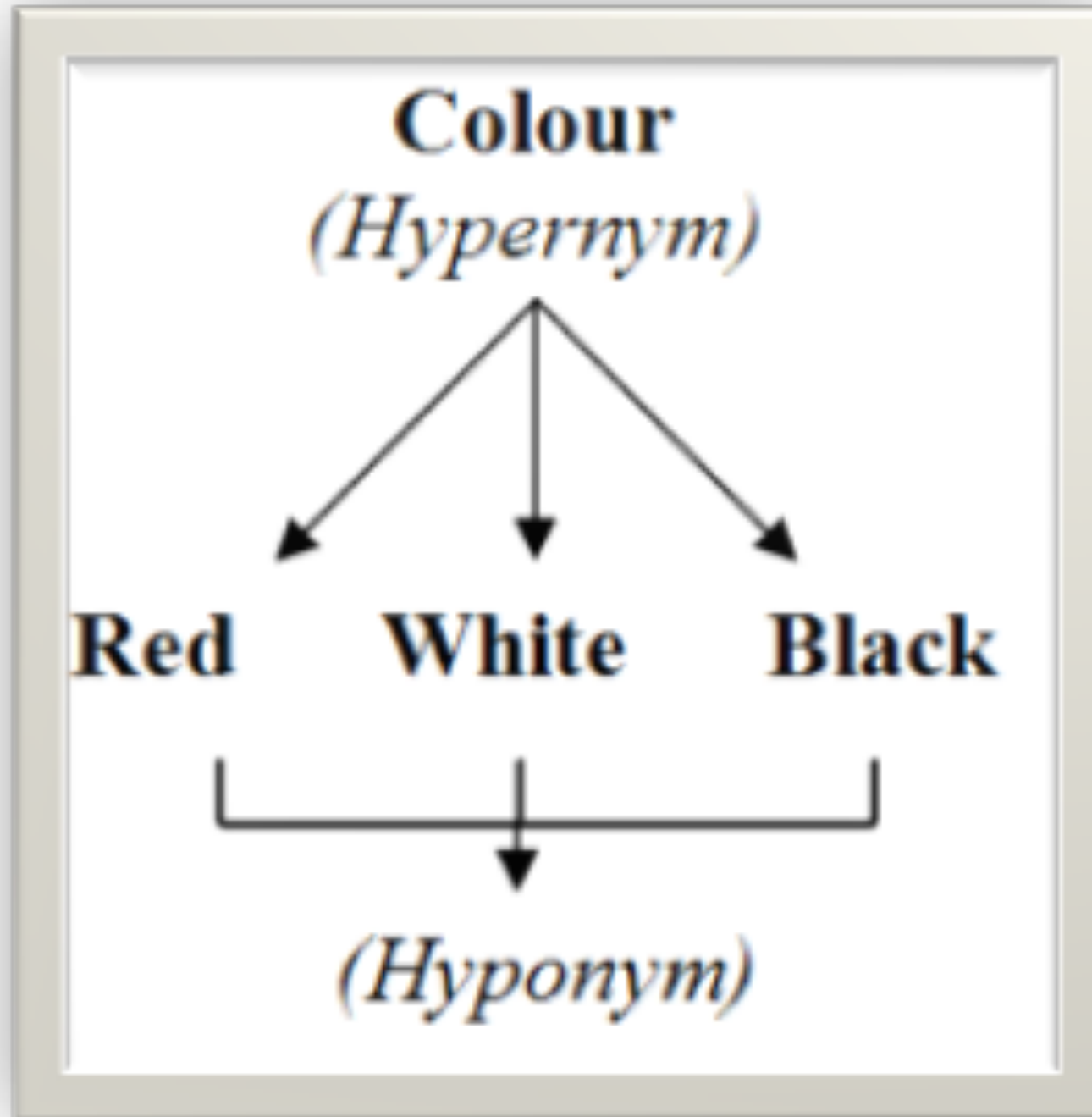
This workshop

- Words and senses
- Wordnet and lexical semantics
- Distributional semantics and word embedding

Senses



Senses



Senses

- **Meronym:** Part of a whole



- **Holonym:** The whole to which parts belong



WordNet

WordNet Search - 3.1

- [WordNet home page](#) - [Glossary](#) - [Help](#)

Word to search for:

Display Options:

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations

Display options for sense: (gloss) "an example sentence"

Noun

- [S:](#) (n) **information**, [info](#) (a message received and understood)
 - [direct hyponym](#) / [full hyponym](#)
 - [direct hypernym](#) / [inherited hypernym](#) / [sister term](#)
 - [S:](#) (n) [message](#), [content](#), [subject matter](#), [substance](#) (what a communication that is about something is about)
 - [S:](#) (n) [communication](#) (something that is communicated by or to or between people or groups)
 - [S:](#) (n) [abstraction](#), [abstract entity](#) (a general concept formed by extracting common features from specific examples)
 - [S:](#) (n) [entity](#) (that which is perceived or known or inferred to have its own distinct existence (living or nonliving))
 - [derivationally related form](#)
- [S:](#) (n) **information** (knowledge acquired through study or experience or instruction)
- [S:](#) (n) **information** (formal accusation of a crime)
- [S:](#) (n) [data](#), **information** (a collection of facts from which conclusions may be drawn) "statistical data"
- [S:](#) (n) **information**, [selective information](#), [entropy](#) ((communication theory) a numerical measure of the uncertainty of an outcome) "the signal contained thousands of bits of information"

entity abstraction... communication message...	entity abstraction... psychological... cognition...	entity abstraction... communication message... statement pleading charge... accusation...	entity abstraction... group... collection...	entity abstraction... measure system of meas... information meas...
information				
entity physical... process... processing data process... operation computer op...	entity abstraction... psychological... cognition... process... basic cog... memory...	entity abstraction... psychological... cognition... process... basic cog... memory...	entity abstraction... psychological... event act...	
retrieval				

information is more similar to the word *retrieval* or the word *science*

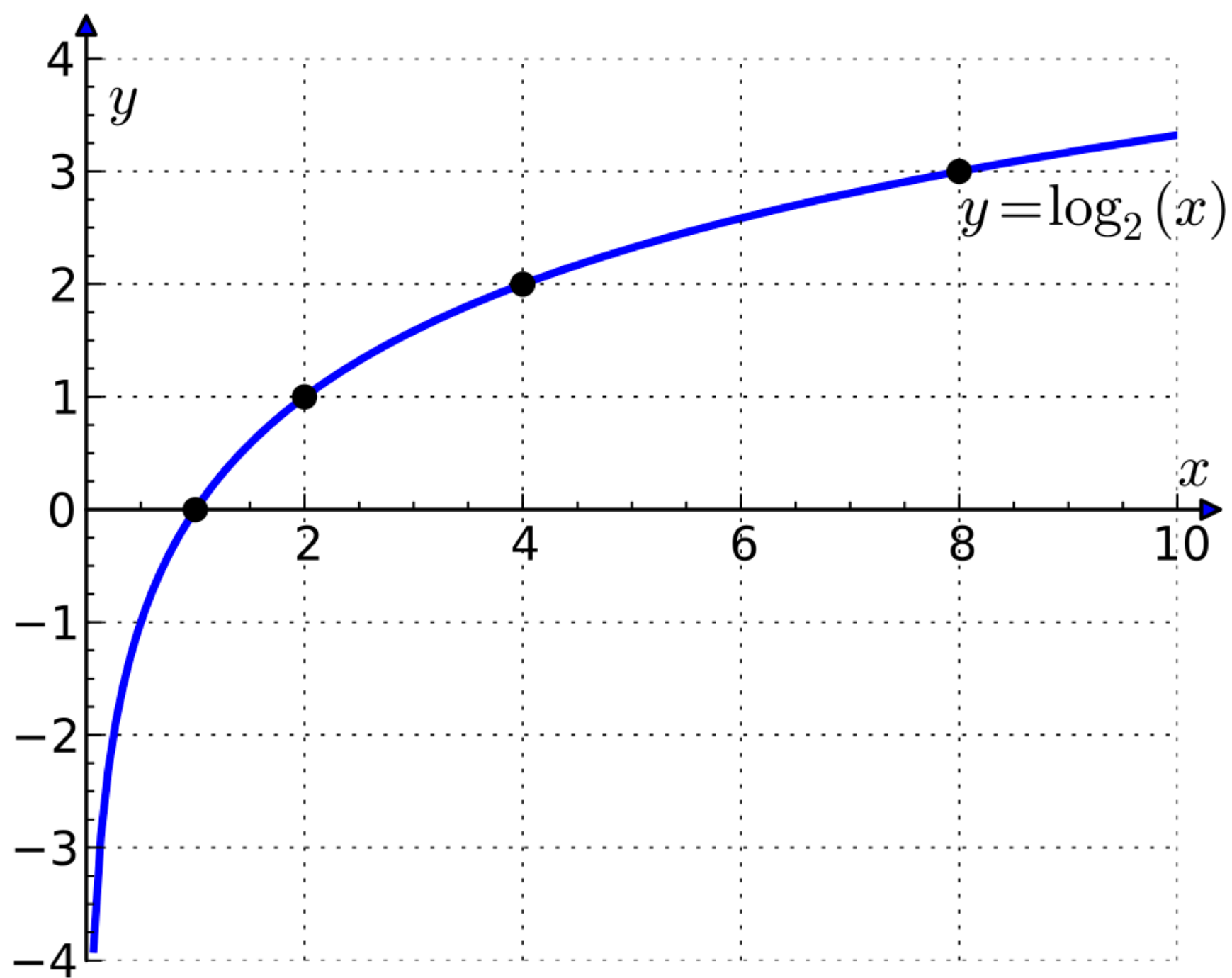
$$WuP_sim(c_1, c_2) = \frac{2 \times depth(LCS(c_1, c_2))}{depth(c_1) + depth(c_2)}$$

		information				
retrieval		1	2	3	4	5
	1	0.154	0.154	0.118	0.154	0.143
	2	0.308	0.615	0.235	0.308	0.286
	3	0.364	0.545	0.267	0.364	0.333

Q4a PMI

	cup	not (cup)	Total
world	55	225	280
not (world)	315	1405	1720
Total	370	1630	2000

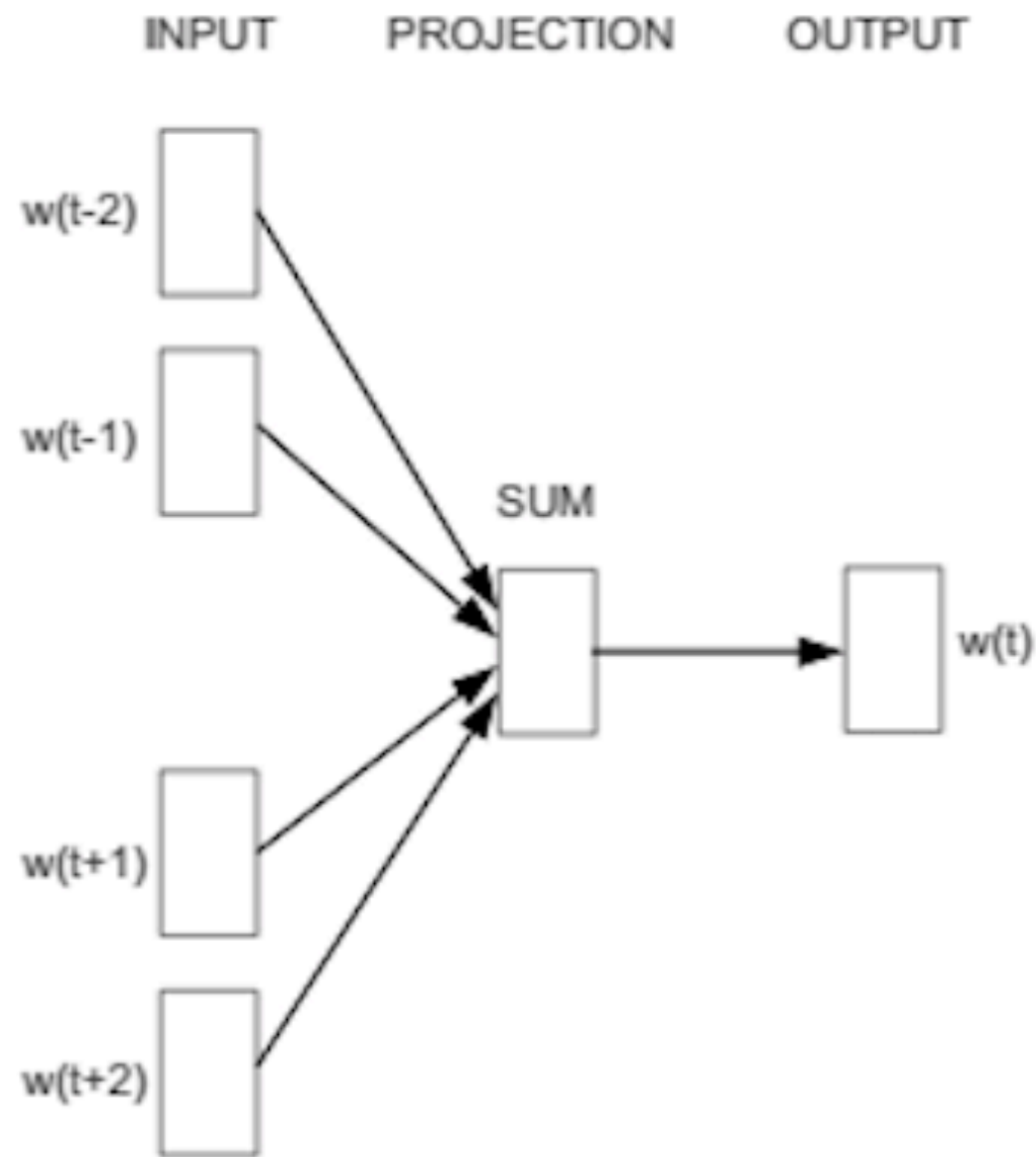
$$PMI(x, y) = \log_2 \frac{p(x, y)}{p(x)p(y)} = \log_2 P(x, y) - \log_2 p(x) - \log_2 P(y)$$



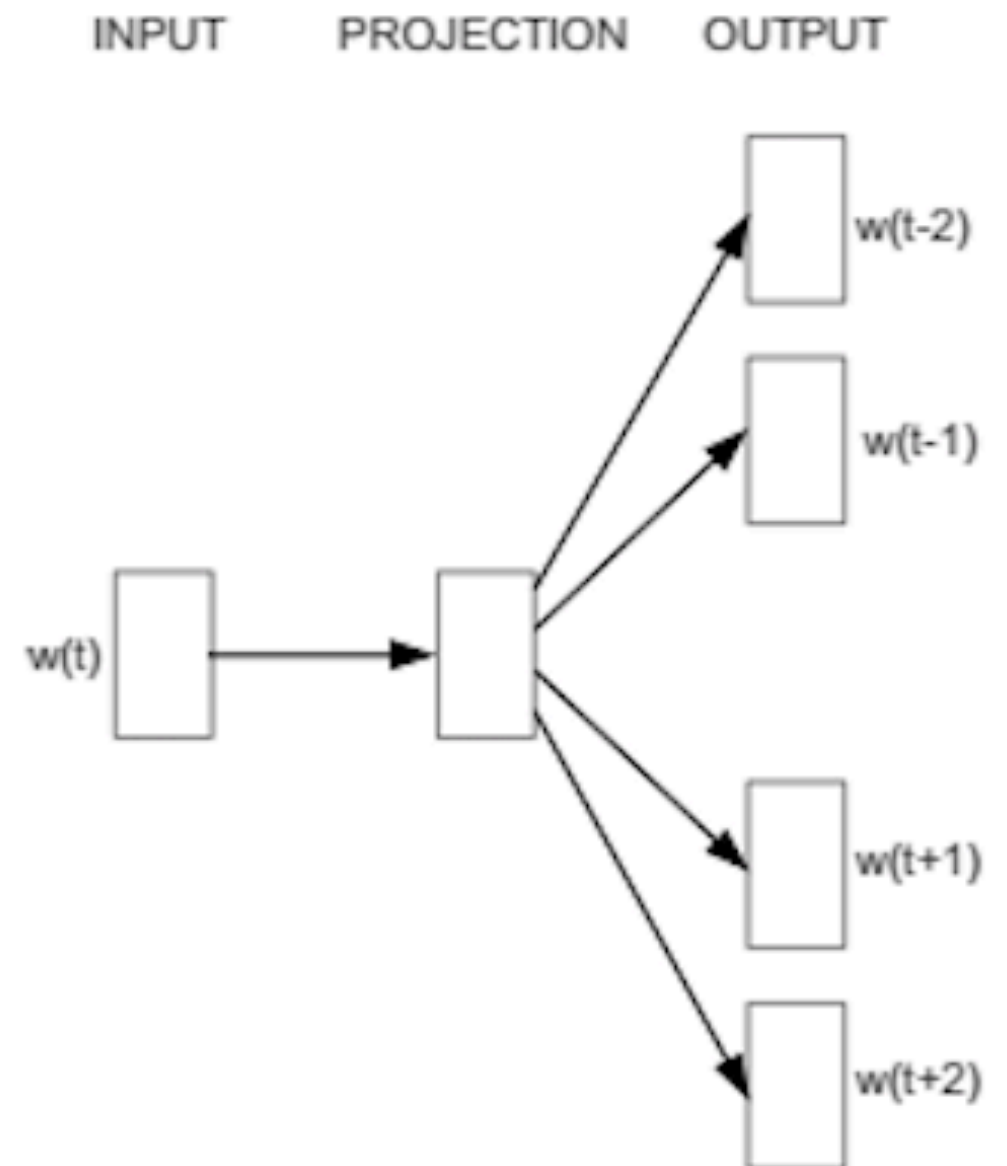
$$PMI(x, y) = \log_2 \frac{p(x, y)}{p(x)p(y)} = \log_2 \frac{p(y|x)}{p(y)} = \log_2 \frac{p(x|y)}{x}$$

Q6 word to vector

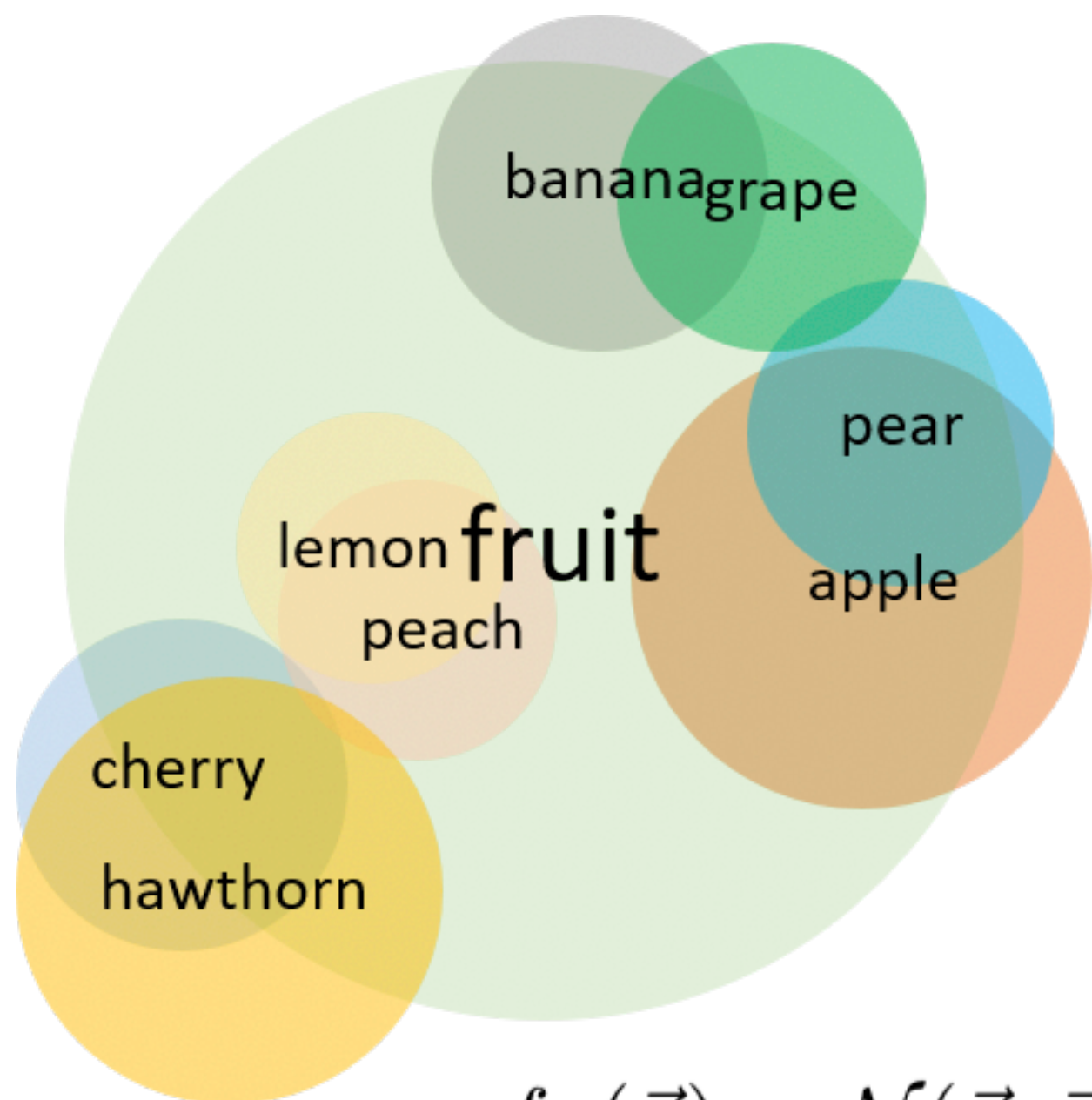
<https://skymind.ai/wiki/word2vec>



CBOW



Skip-gram



Gaussian Word Embedding

$$\begin{aligned} f_w(\vec{x}) &= \mathcal{N}(\vec{x}; \vec{\mu}_w, \Sigma_w) \\ &= \frac{1}{\sqrt{(2\pi)^D |\Sigma_w|}} e^{-\frac{1}{2}(\vec{x} - \vec{\mu}_w)^\top \Sigma_w^{-1} (\vec{x} - \vec{\mu}_w)} \end{aligned}$$