

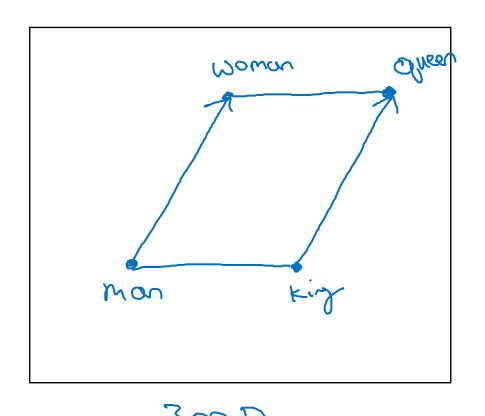
# NLP and Word Embeddings

# Properties of word embeddings

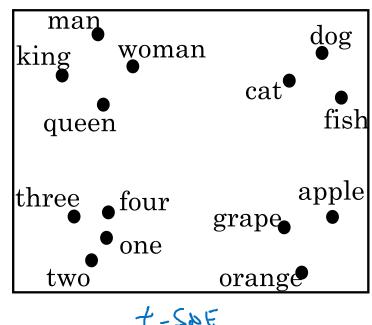
## Analogies

	Man (5391)	Woman (9853)	King (4914)	Queen (7157)	Apple (456)	Orange (6257)
Gender	-1		-0.95	0.97	0.00	0.01
Royal	0.01	0.02	0.93	0.95	-0.01	0.00
Age	0.03	0.02	0.70	0.69	0.03	-0.02
Food	0.09	0.01	0.02	0.01	0.95	0.97
	Q 5391 Q man	2 woman	2 0	eman - e	$\omega$ $\approx$ $\begin{bmatrix} -2 \\ 0 \\ 0 \end{bmatrix}$	
Mon -> Woman & King ->? Queen  Ching - Cyman N Cking - C?  Lane  Lane  Ching - Cyman N Cking - C?  Lane  Lane  Ching - Cyman N Cking - C?						
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#### Analogies using word vectors







t-SAF

$$e_{man} - e_{woman} \approx e_{king} - e_{woman} \approx e_{woman} \approx e_{king} - e_{woman} \approx e_{woman} \approx e_{woman} = e_{woman} \approx e_{woman} = e_$$

300 D Sim (Qw, Pking - eman + ewoman) Find word wi arg mox

Andrew Ng

### Cosine similarity

$$\Rightarrow sim(e_w, e_{king} - e_{man} + e_{woman})$$

$$Sim(u,v) = \frac{u^{T}v}{\|u\|_{2}\|v\|_{h}}$$

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$$|u-v||$$

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Man:Woman as Boy:Girl

Ottawa:Canada as Nairobi:Kenya

Big:Bigger as Tall:Taller

Yen:Japan as Ruble:Russia