## Scaling up language technology for the next 1000 languages

Kevin Scannell
Saint Louis University
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## Linguistic Diversity: The Numbers

- About 7100 languages spoken in the world (Ethnologue)
- Almost half are "endangered" (UNESCO)
- 2500-3000 have some online presence
- < 1000 written by the language community
- Wikipedias for about 300 languages
- Tweets in about 275 languages
- Spell checkers for about 180
- Google search interface translated into 150
- Google Translate in about 100


## Language Technology

- Machine translation
- Speech recognition
- Predictive text
- Search engines
- Dialogue systems
- Spelling/grammar checking
- Text normalization (e.g. modernization)
- Optical character recognition
- Handwriting recognition


## Example: Context-Sensitive Spell checking

- In your brain: B = "May I borrow your phone?"
- On your screen: S = "May I borrow yore phone?"
- Think of S as a "distortion" of B
- Job of a spell checker is to "decode" the original message B from S
- For simplicity assume we know all other words are correct
- Let $\mathrm{S}(w)$ be the result of substituting word $w$ for "yore"
- Want to find the w maximizing P(S(w) I S)
- Bayes' Law, same as maximizing $P(S \mid S(w)) \times P(S(w))$
- Model first term as P(yore I w); think of w = yore, your, tore, pen, etc.
- Second term is simply the probability of the sentence: "Language Model"


## Noisy Channels

- Almost all of the language technologies I mentioned can be modeled this way
- Machine translation, for example, from French to English
- Weird perspective on the French language!
- Given French sentence F, pick English sentence E maximizing P(F I E) $\times P(E)$
- $P(F \mid E)$ is the "translation model"; often word or phrase-based models
- Different channel (speech, OCR, etc.) => different channel model
- Second term is the same across all of these problems!
- Better language models => better language technology
- (Incidentally this is why translation into English is usually better...)


## Language modeling

- Chomsky: "... the notion of 'probability of a sentence' is an entirely useless one, under any known interpretation of this term."
- Let $S=W_{1} W_{2} W_{3} \ldots w_{n}$
- $P(S)=P\left(\left.w_{1}\right|^{\wedge}\right) P\left(w_{2} \mid w_{1}\right) P\left(w_{3} \mid w_{1} w_{2}\right) \ldots P\left(w_{n} \mid w_{1} w_{2} w_{3} \ldots w_{n-1}\right)$
- Usually formulated and computed this way (word prob. given history)
- Humans are pretty good at estimating these, at least
- $\quad \mathrm{P}$ (FridaylMy party is this coming) $>\mathrm{P}$ (TuesdaylMy party is this coming)
- $\quad \mathrm{P}$ (is IThe man with the glasses) $>\mathrm{P}$ (arelThe man with the glasses)


## Turing Test

- Language models can be used generatively also
- Assume you've already generated $w_{1} w_{2} w_{3} \ldots w_{n-1}$
- Simply choose the next word from the distribution $P\left(w_{n} \mid w_{1} w_{2} w_{3} \ldots w_{n-1}\right)$
- Using full history of a conversation, generate "probable" responses
- Good enough language model => indistinguishable from a human (?)
- Language modeling == Al? (if you believe Turing's formulation)


## N -gram language models

- Markov assumption: $P\left(w_{n} \mid w_{1} w_{2} w_{3} \ldots w_{n-1}\right)=P\left(w_{n} \mid w_{n-k+1} \ldots w_{n-1}\right)$
- Often $k=3$ (trigram modeling): need to compute $P\left(w_{n} \mid w_{n-2} w_{n-1}\right)$
- Trained using large text corpora and counting trigrams
- Smooth models by backing off to lower-order n-grams
- $P($ Fridaylthis coming) >? P(Tuesdaylthis coming)
- P (islthe glasses) < P (arelthe glasses)
- For English, trainable using Google’s 5-gram dataset (trillion word corpus)


## Intrinsic and extrinsic evaluation

- Extrinsic: incorporate in language tech and evaluate end-to-end
- Intrinsic: what probability does the model assign to a big test corpus?
- $S=w_{1} w_{2} w_{3} \ldots w_{N}$ where $N$ is in the millions or more
- Average log-prob of over words (units are bits/word)

$$
\left[-\Sigma \log _{2} P\left(w_{j} \mid w_{1} w_{2} w_{3} \ldots w_{j-1}\right)\right] / N
$$

- Approximation of cross-entropy between language and the model
- Best n-gram models for English just over 6 bits/word
- State-of-the-art neural models under 5 bits/word



## Issues for under-resourced languages

- Pros and cons of language independence
- Lack for large text corpora for training and evaluating models


## "Praistriúchán"

New Irish portmanteau word: "praiseach" = "a mess, a botch job", "aistriúchán" = "translation"


## Incorporating linguistic knowledge

- Celtic languages have "initial mutations"
- Almost always predictable from previous two words
- bád seoil "sailboat", mo bhád seoil "my sailboat", ár mbád seoil "our sailboat"
- N-gram models don’t "see" that these are all the same word
- If most training examples are first type, say, harder to predict collocation
- (Google even gets this wrong in previous image: tríd an mbóthar)
- Easy enough to use a factored language model to get better results for Irish
- Examples like this abound...


## Language models via web corpora

- Crúbadán project; web crawled corpora for under-resourced languages
- Began around 2003 for the six Celtic languages
- Now crawling 2225 languages, hundreds more queued for training
- Scaled up thanks to NSF grant 1159174 (Linguistics)
- http://crubadan.org/
- For Irish, more than 100 million words (> 175 million with offline material)



## Unlocking the past: fonts and manuscripts

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## Unlocking the past: standardization

> Gaeilge Gaedhilge Gaeilige Gaelige Gailge Gaoidhilge Gaoidheilge Gaelge Gaidhlige Gaedheilge Gaoidhelge Gailege Gaielge Gaodhailge Gaedilge Gaeidhilge Gaedhilige Gaoidilge Gaeilgele Gaedhlige Gaédhilge Gaoilge Gaeillge Gaeilga Gaidhilge Gaelilge Gaodheilge Gaeiilge Gaedhilghe Gadhilge Gaheilge Gaellge Gaoilaige Gaodhilge Gaedhilgé Gaeilege Gaaeilge Gailige Gaeilgé Gaeghilge Gaedhailge Gaoidhlige Gaelgie Gaeiloge Gaeilgle Gaeilghe Gaellge Gaeidhlge Gaeidheilge Gaeeilge Gaoilige Gaóilge Gaoilaga Gaoigheilge Gaoidhlge Gaoidelge Gaoideilge Gaodhéilge Gaieilge Gaeulge Gaeuilge Gaeolge Gaeoidheilge Gaeilgi Gaeilgee Gaeílge Gaeidlge Gaeidilge Gaeidhelge Gaehilge Gaeeilgee Gaedhlge Gaedhiilge Gaedhelga Gaédhailge Gaedgilge Gadehilge Gaddhilge Gaoghailge Gaileige Gaidlhlige Gaidhlge Gaeliage Gaelga Gaéilge Gaedilghe Gaedhulge Gaedhalg Gaedheilg Gaédheilg Gaedhilg Gaedhilig Gaeilg Gaoidhealg Geadhilge Geailge

## Royal Irish Academy Corpus

corpas.ria.ie - 19 million words digitized and indexed by standard forms


