

Neural Polysynthetic Language Modelling





**2019 | INTERNATIONAL YEAR OF
Indigenous Languages**



Background

Intersecting machine learning & linguistic fieldwork

St. Lawrence Island Yupik

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- Increasing understanding, reconciliation and international cooperation.
 - Creation of favorable conditions for knowledge-sharing & dissemination of good practices with regards to indigenous languages.
 - Integration of indigenous languages into standard setting.
 - Empowerment through capacity building.
 - Growth and development through elaboration of new knowledge.

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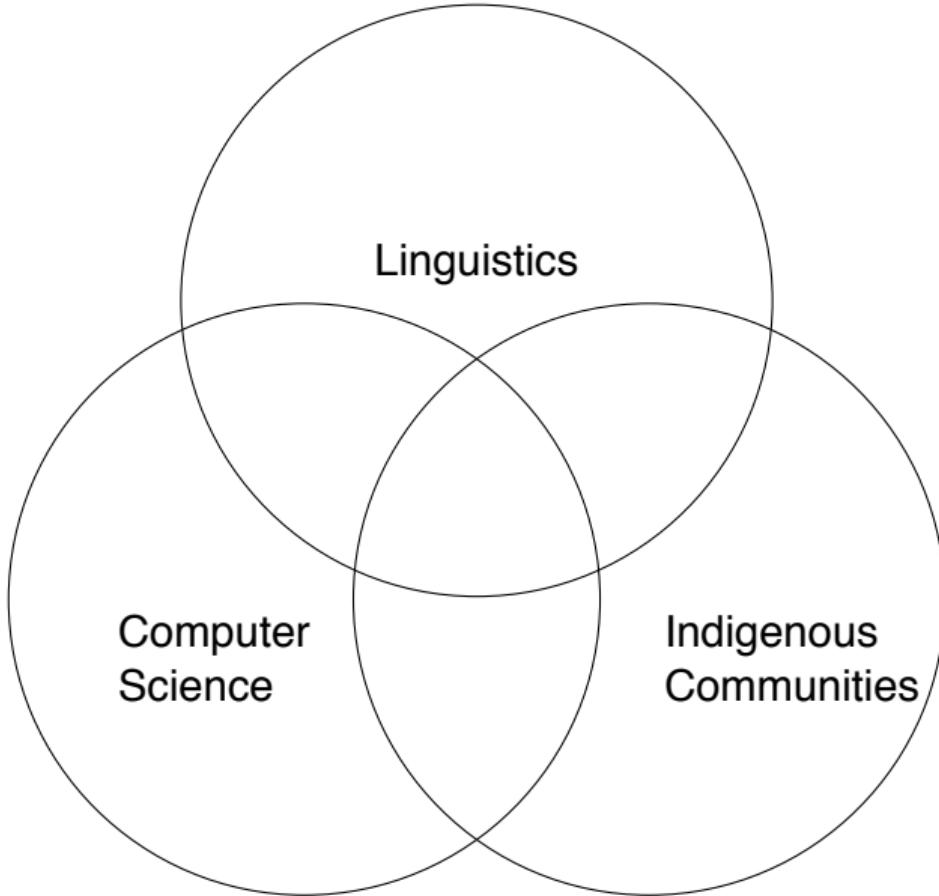
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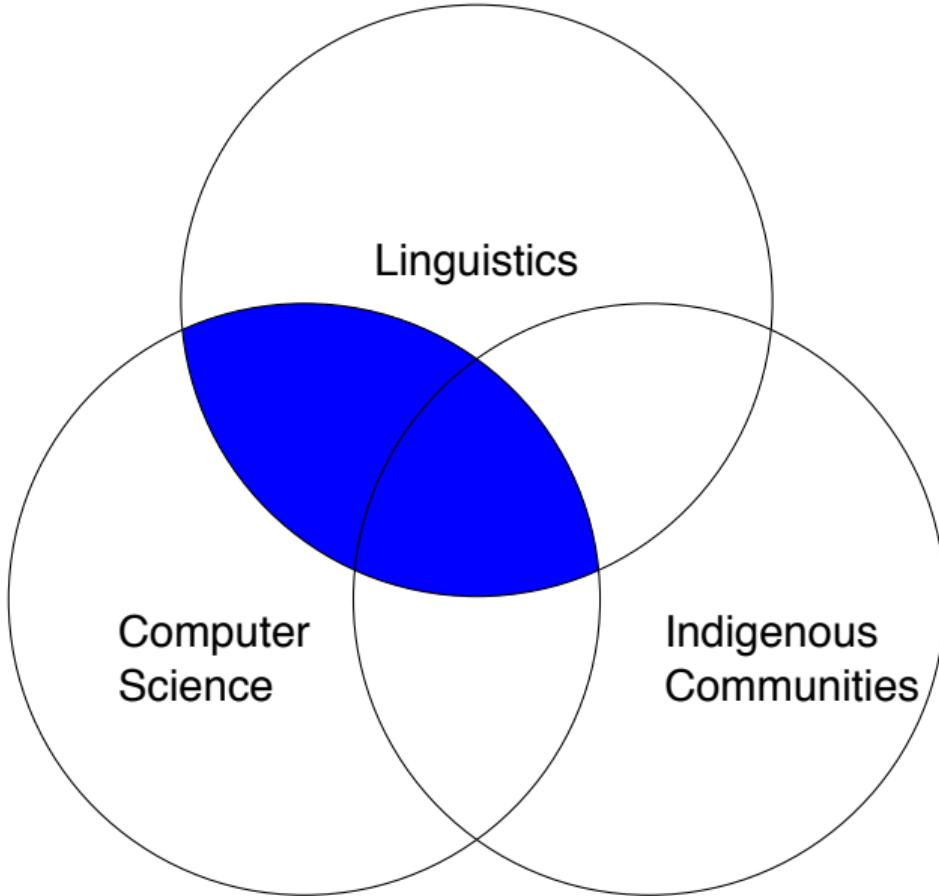
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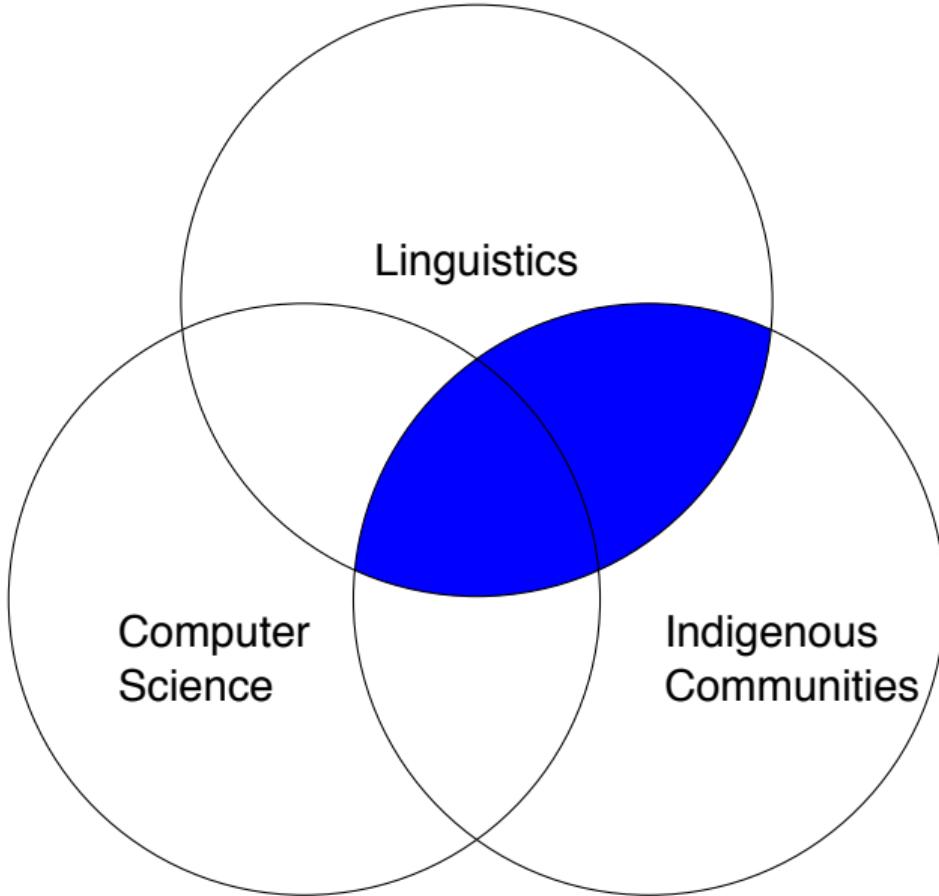
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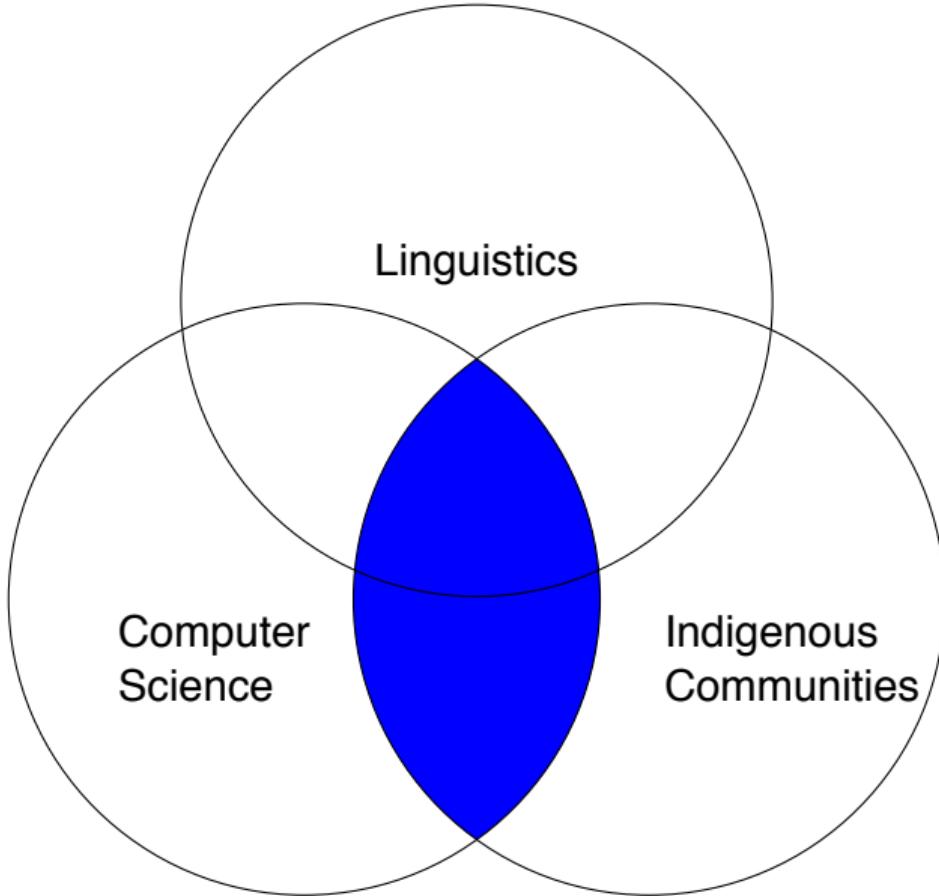
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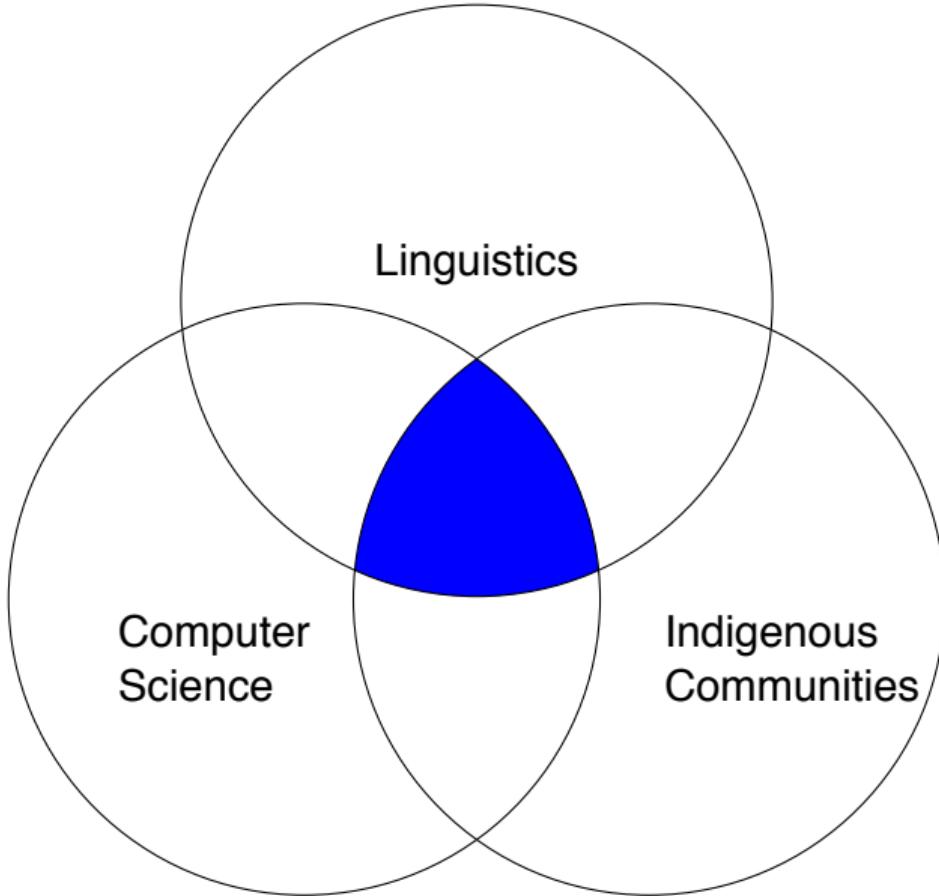
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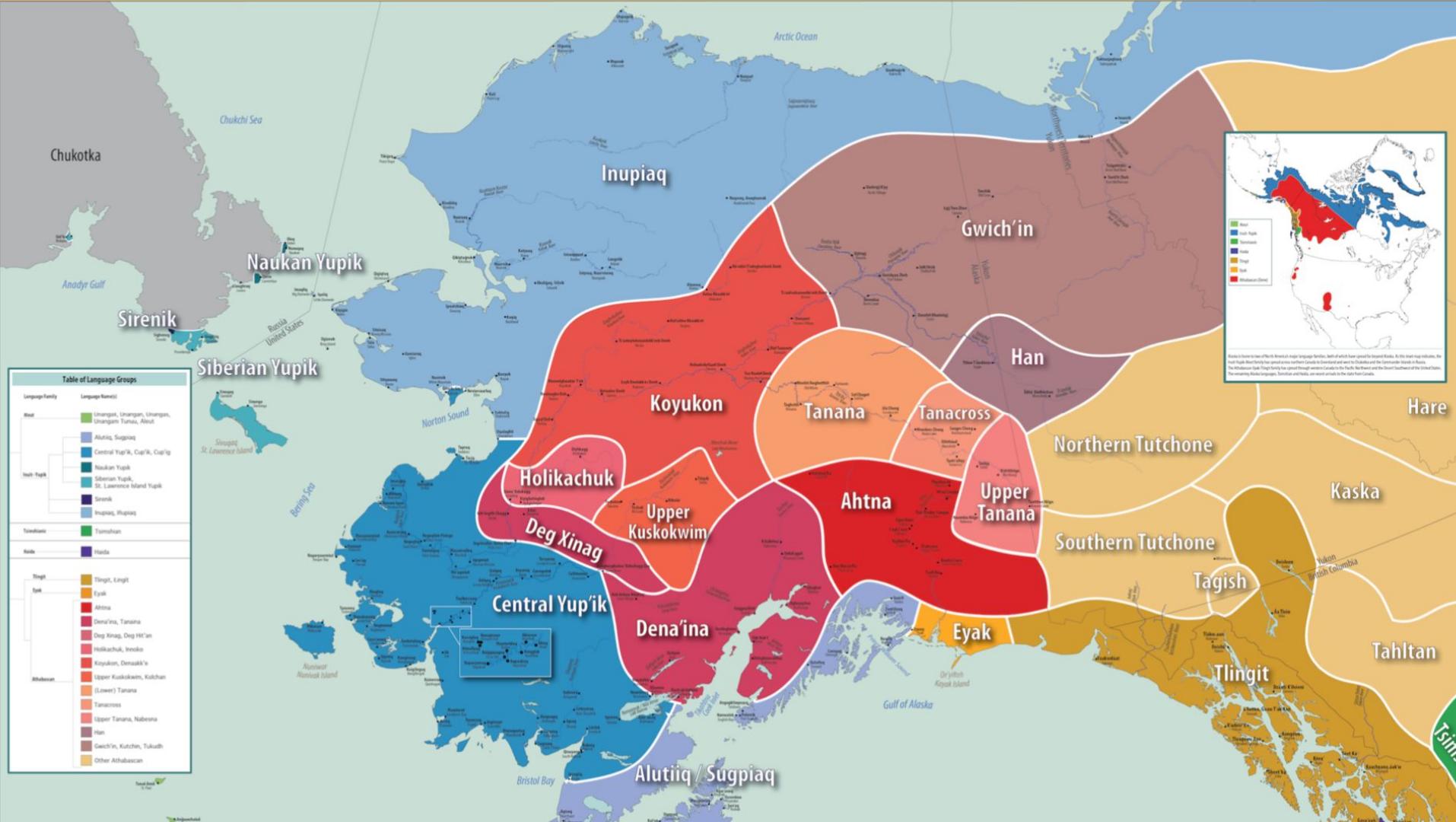
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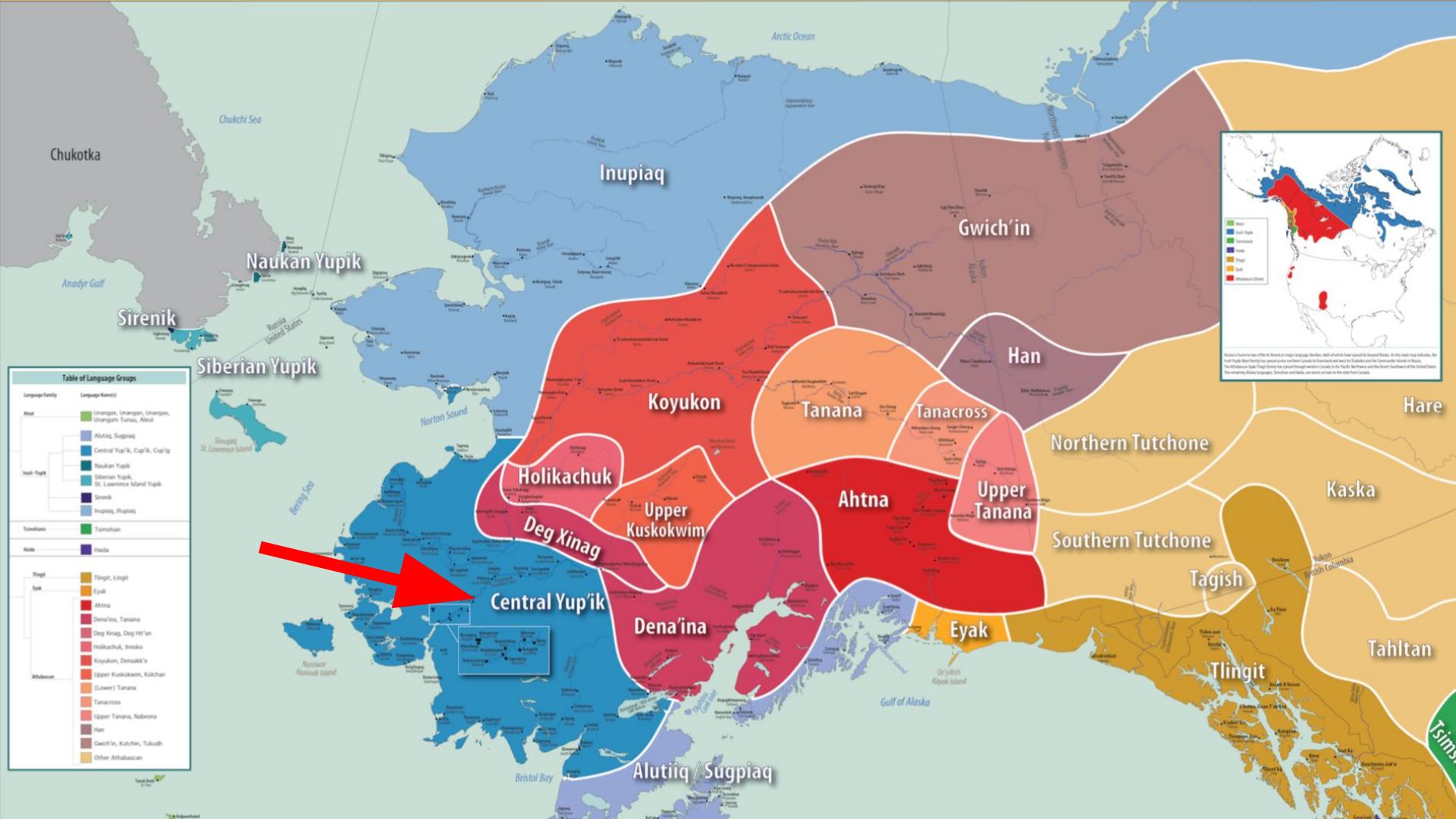
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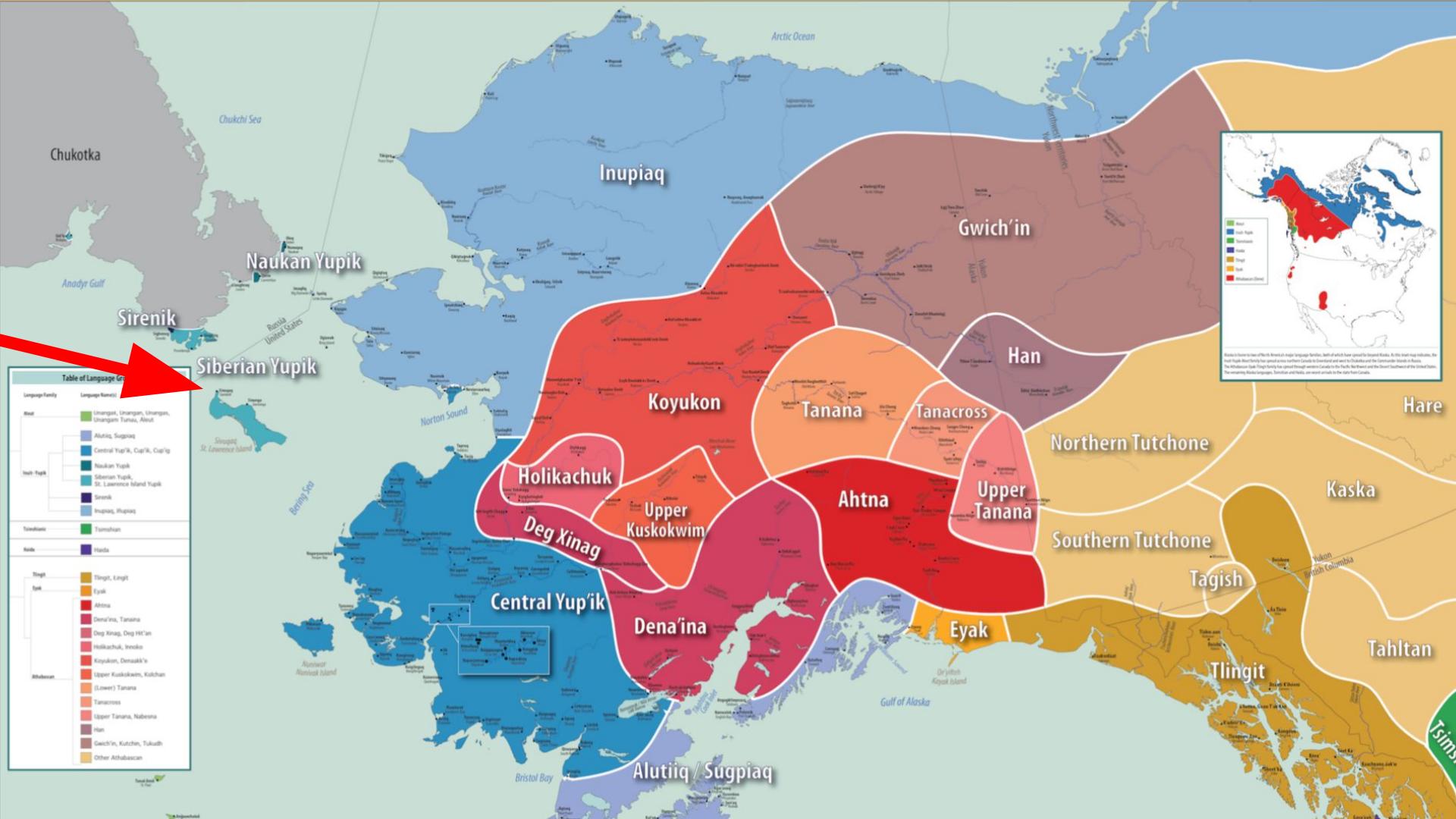
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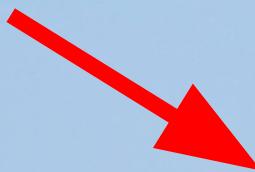












Inuit-Yupik-Unangam Tunuu language family

- Greenland (Inuit)
- Northern Canada (Inuit)
- Northern Alaska (Inuit)
- Western Alaska (Yup'ik)
- Southwestern Alaska (Sugpiak, Unangam Tunuu)
- **St. Lawrence Island (Yupik)**
- Big Diomede (Inuit)
- Far eastern Russia (Yupik, Sirenik)

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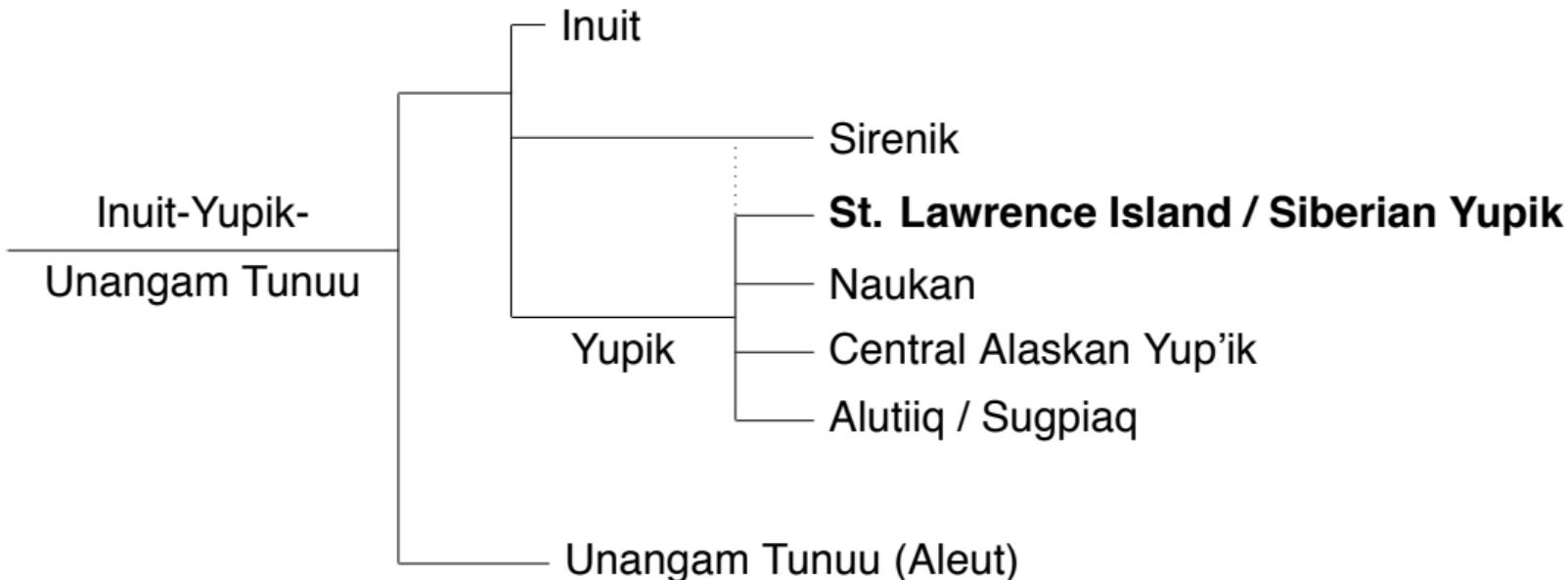
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Inuit-Yupik-Unangam Tunuu language family



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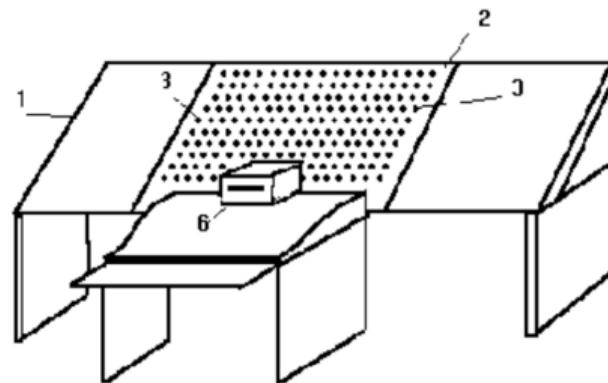
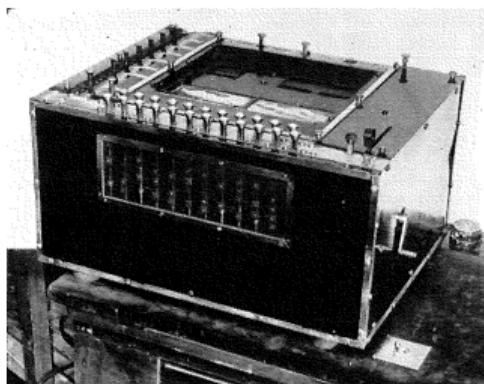
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Since 1933, NLP technology has overwhelmingly focused on languages & methodologies in which the word is the primary meaning-bearing unit



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For *most* human languages, this assumption is
fundamentally broken



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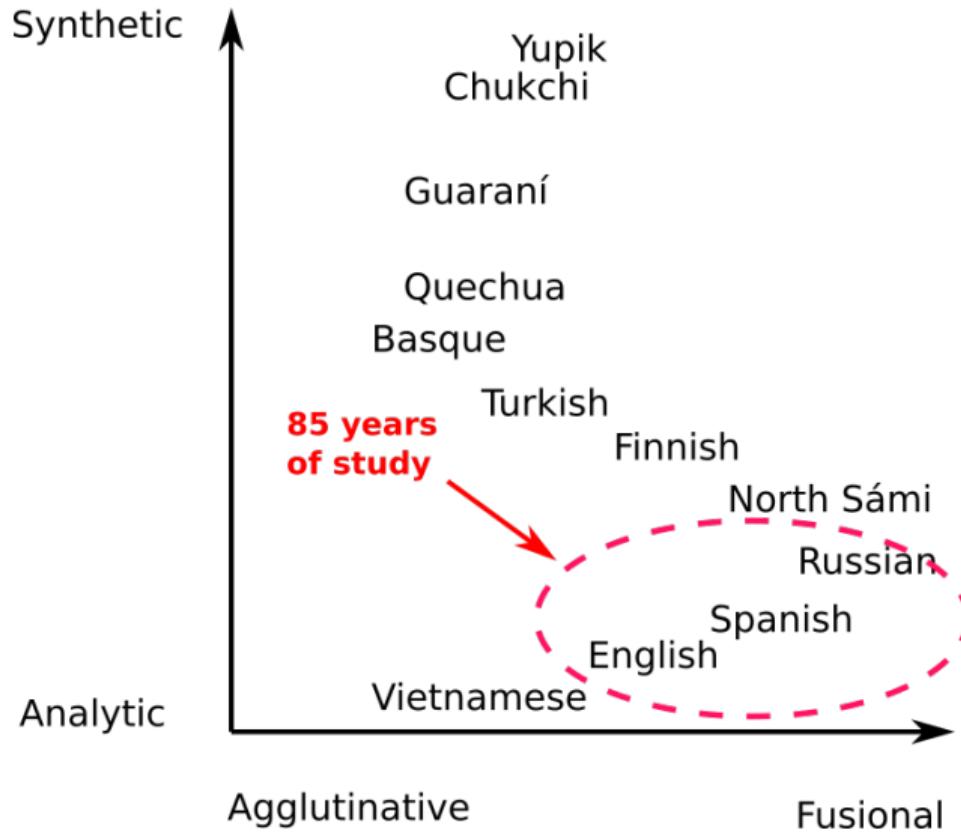
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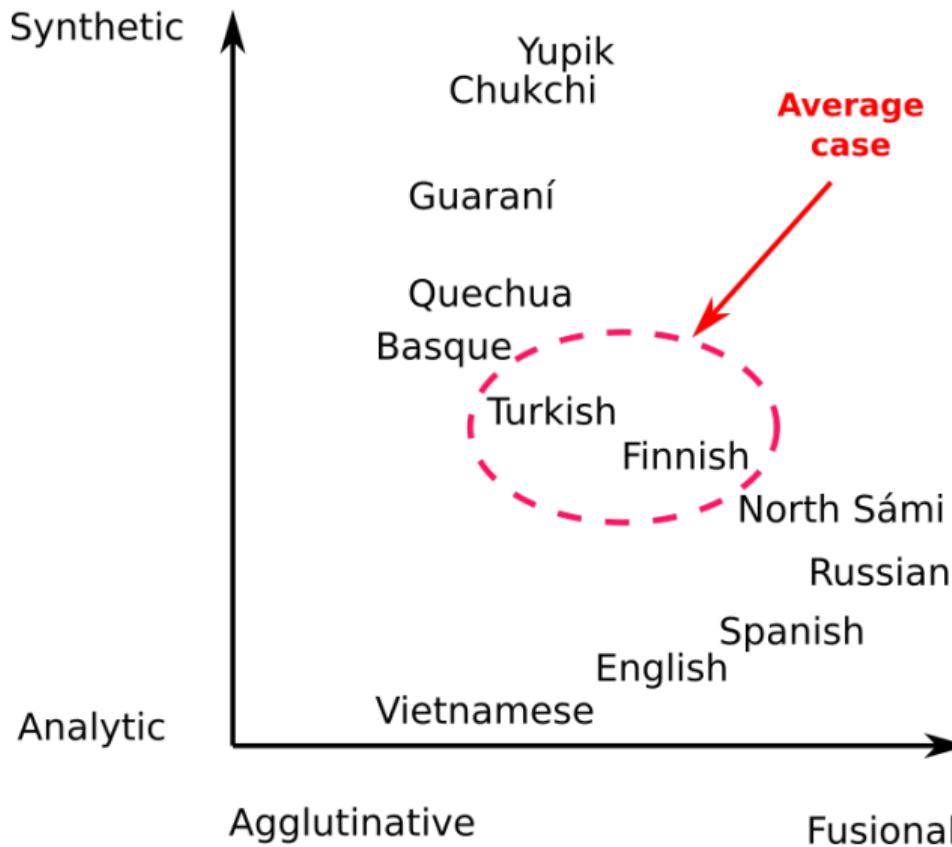
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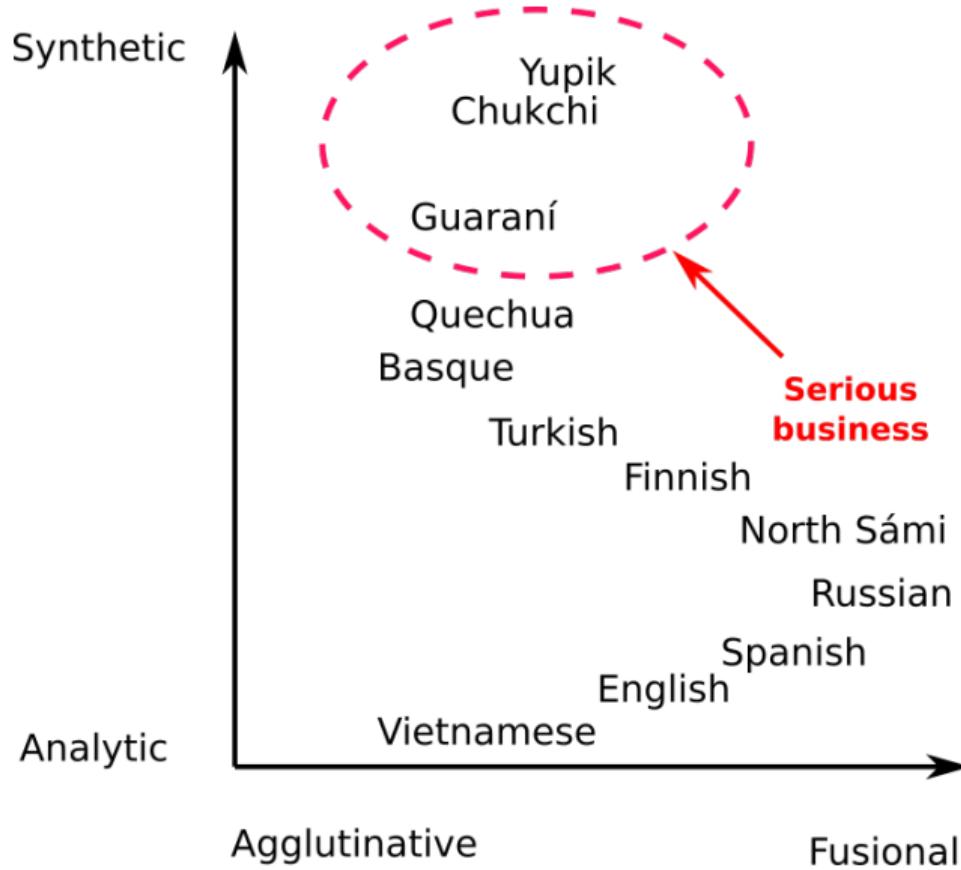
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$$p(\tau_t | \tau_1 \dots \tau_{t-1})$$



$$p(\tau_t | \tau_1 \dots \tau_{t-1}) = \frac{\text{count}(\tau_1 \dots \tau_t)}{\text{count}(\tau_1 \dots \tau_{t-1})}$$



English



* actual data disparity
is much much larger

dog

dogs

qikmiq

qikmik

qikmit

qikmiq

qikmik

qikmit

qikmii

qikmikek

qikmiik

qikmigka

qikmigken

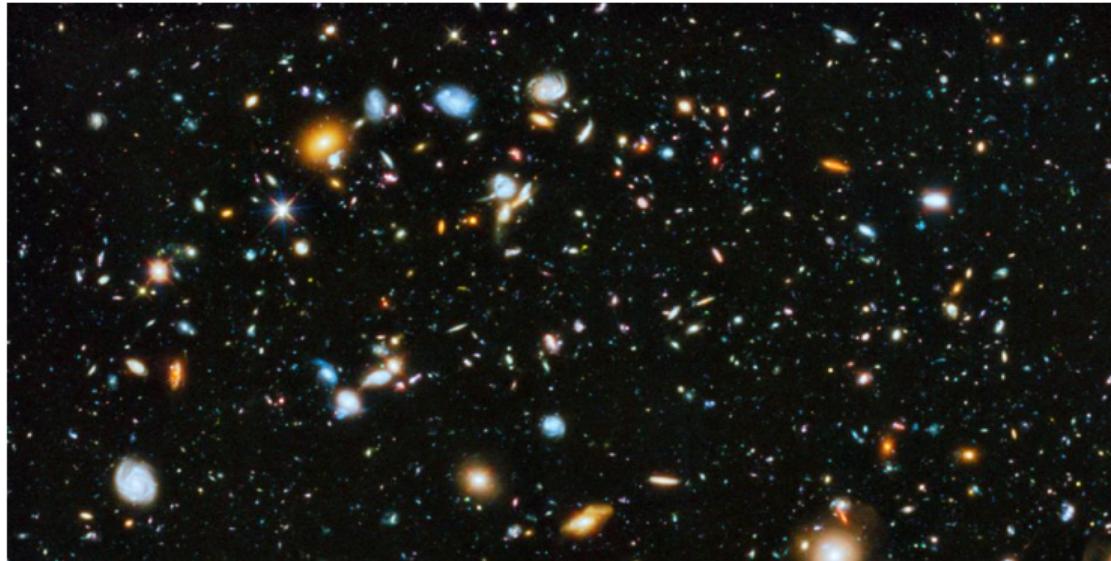
qikminka

qikmiqa

qikmighpung

qikmighput

There are 1.2×10^{23} stars in the observable universe.



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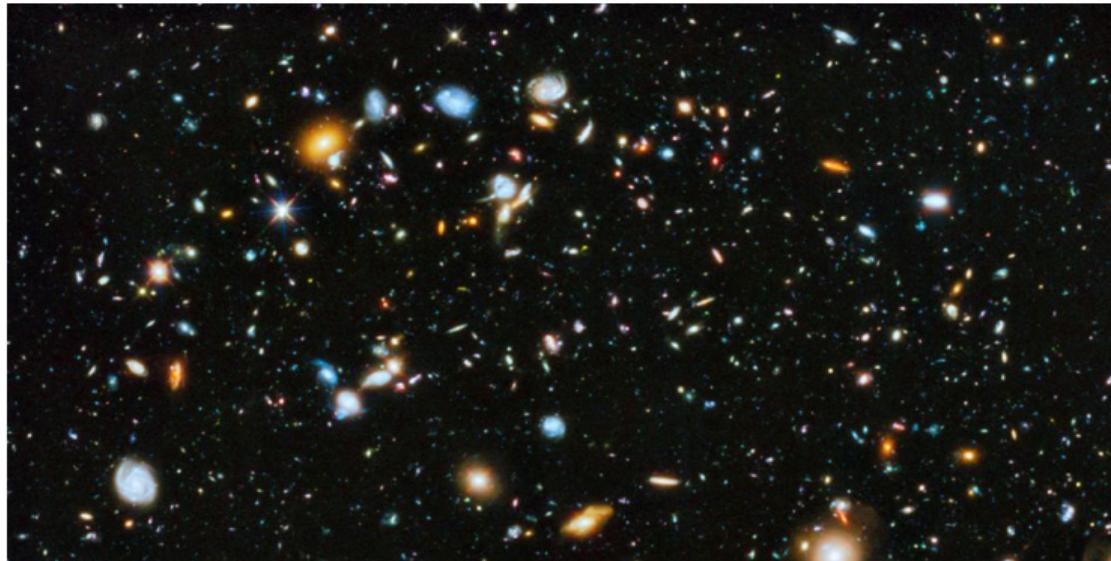
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There are 1.2×10^{23} possible Yupik word forms.



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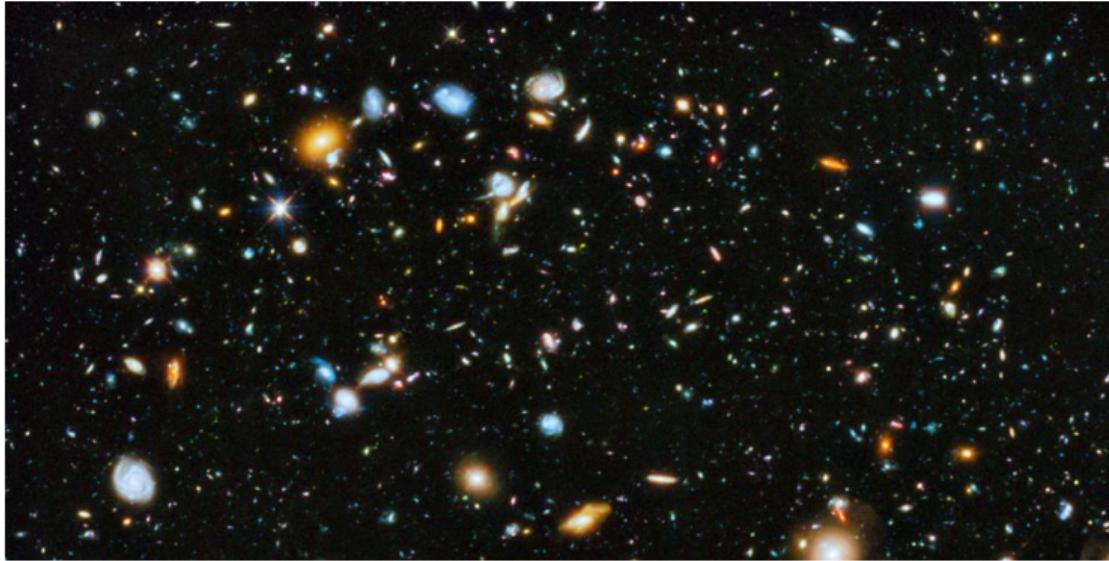
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Big data is NOT the solution.



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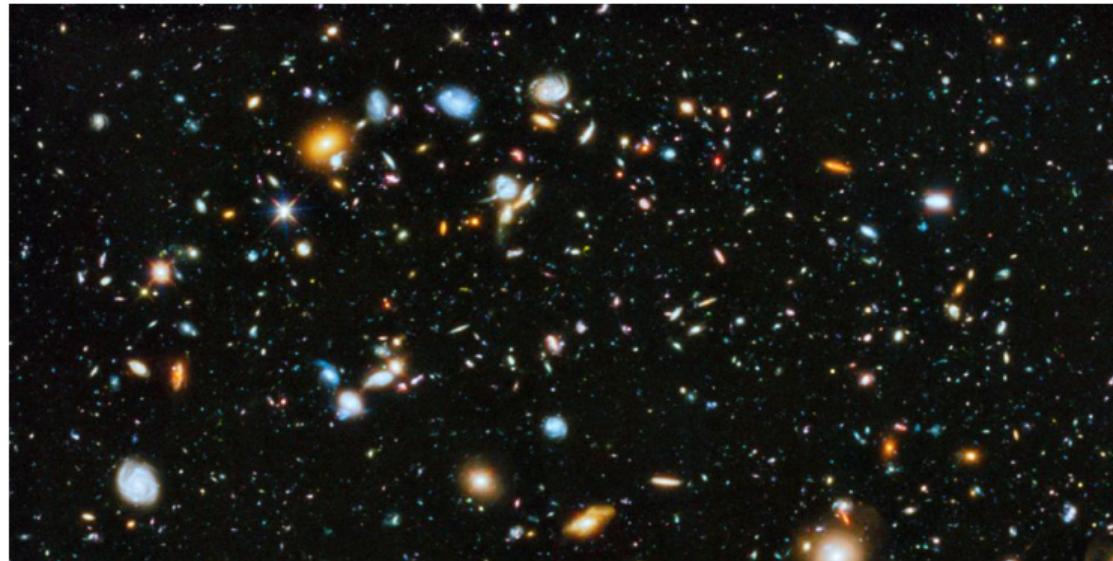
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Modelling only at the word-level is like modelling only at a galaxy-level.



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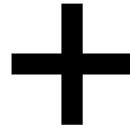
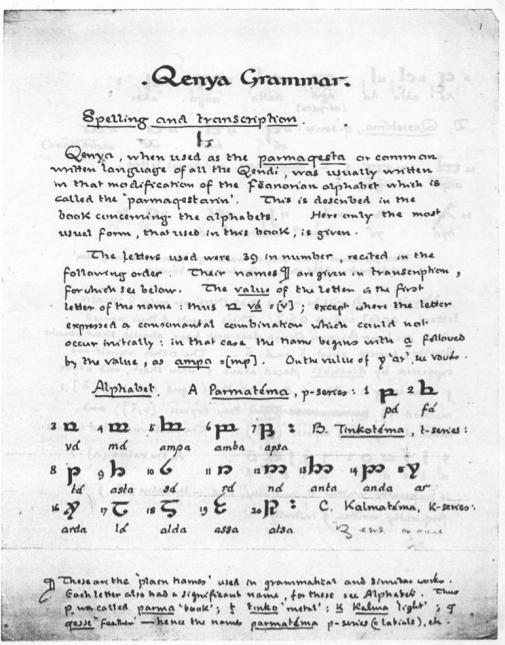
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“**ქართული ენა**”
“**ქართული ენა**”
“**ქართული ენა**”
“**ქართული ენა**”





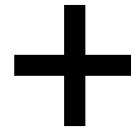
PARMA ELOALAMBERON XXII

The Feanorian Alphabet · PART I

Quenya Verb Structure

by J. R. R. TOLKIEN

ᛇᚠᚢጀ ᛉጀ ᛈጀ ᛉጀ
 ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ ᛉጀ



ash	nazg	durb-	-at-	-ul-	-ûk
one	ring	to.dominate	[Ptcp]	[3PI]	[Compl]



...

agh	burzum	ishi	krimp-	-at-	-ul
and	darkness	inside	to.bind	[Ptcp]	[3PI]

Course goals

- Learn about a new language from a reference grammar
 - Demonstrate your understanding through writing and teaching
- Select a topic from computational linguistics applicable to this language
 - Conduct a literature review, resulting in an annotated bibliography & report on state of the art
- Perform research on this topic
 - Identify state-of-the-art baseline, implement & extend it, run experiments, write a paper
- Conduct extended research in a group
 - Collaborate, experiment, and jointly author a paper
- Act as a peer reviewer for your classmates' work

Demographics

- 1300 Yupiget on St. Lawrence Island
- 800 Yupiget on Russian mainland
- 300-400 Yupiget on Alaskan mainland

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Education

- 1930s-1950s Yupik materials developed in Russia
- 1970s-1990s Yupik materials developed in Alaska

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Language shift - Russia

- By mid-20th century, shift away from Yupik in Russia
- Current estimate of < 200 L1 Yupik speakers in Russia
- Youngest L1 Russian Yupiget estimated age > 70

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Language shift - St. Lawrence Island

- In 1980, nearly all St. Lawrence Island Yupiget children spoke Yupik at home
- By mid-1990s through early 2000s, shift away from Yupik among SLI youth
- All SLI Yupiget born 1980 or earlier assumed to be L1 Yupik
- Current estimate of at least 540 L1 Yupik speakers on SLI
- Youngest L1 SLI Yupiget not known

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Phonology & Orthography

Close Vowels	i i и	u u ы	Latin IPA Cyrillic
Mid Vowel		e ə ы	Latin IPA Cyrillic
Open Vowel		a ɑ а	Latin IPA Cyrillic

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Syllable structure

- Word-initial V(C)
 - Otherwise CV(C)
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- V may be short (e, a, i, u) or long (aa, ii, uu)
 - Adjacent consonants only at syllable boundaries
 - Adjacent consonant generally must agree in voicing

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Phonology & Orthography

	Labial	Alveolar	Palatal	Retroflex	Velar	Velar (rounded)	Uvular	Uvular (rounded)	Glottal	
Unvoiced Stops	p	t			k	kw	q	qw		Latin
	p	t			k	k ^w	q	q ^w		IPA
	п	т			к	къ	қ	қъ		Cyrillic
Voiced Continuants	v	l	z	y	r	g	w	gh	ghw	Latin
	v	l	z	j	ɿ	ɣ	ɣ ^w	χ	χ ^w	IPA
	в	л	з	иі	р	(г)ъ	г	гъ	гъ	Cyrillic
Unvoiced Continuants	f	ʃ	s	ʂ	gg	wh	ghh	ghhw	h	Latin
	f	ʃ	s	ʂ	x	x ^w	χ	χ ^w	h	IPA
	ф	љь	с	ш	х	хъ	х	хъ	г	Cyrillic
Voiced Nasals	m	n			ng	ngw				Latin
	m	n			ŋ	ŋ ^w				IPA
	м	н			ң	ңъ				Cyrillic
Unvoiced Nasals	mm	nn			ngng	ngngw				Latin
	мъ	нъ			ңј	ңъ ^w				IPA
	мъ	нъ			ңъ	ңъъ				Cyrillic

Legacy Digitization

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- 3-volume Lore of St. Lawrence Island
- 3-volume Elementary Yupik readers
- 1-volume of Russian Yupik stories

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- Yupik is polysynthetic, allowing for morphologically-complex words

(1) mangteghaghllangllaghyugtukut

mangteghagh-	-ghllag-	-ngllagh-	-yug-	-tu-	-kut
house-	-big-	-build-	-want.to-	-INTR.IND-	-1PL

'We want to build a big house'

- Yupik words typically adhere to the following template:

Root + 0-7 Derivational Morpheme(s) + Inflectional Morphemes + (Enclitic)

- Yupik is polysynthetic, allowing for morphologically-complex words

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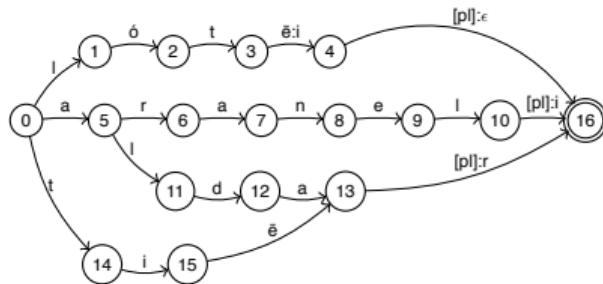
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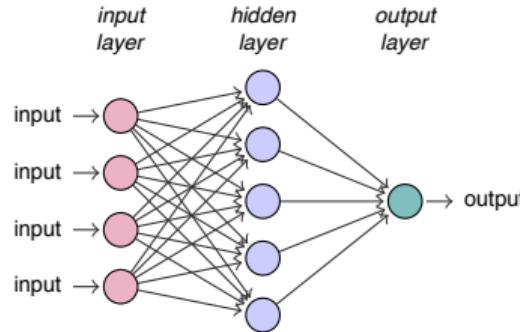
Root + 0-7 Derivational Morpheme(s) + Inflectional Morphemes + (Enclitic)

- Morphological analyzers may be implemented as a

Finite-State Transducer



Neural Network



- Neural systems require LOTS of data
 - But Yupik is a low-resource language
 - Very few surface form-lexical form pairs available

- **OBJECTIVE:** Analyze inflected Yupik nouns with no derivational morphology
 - **TRAINING DATA:** Every nominal surface form and its respective lexical form
 - 3873 Yupik noun roots
 - 273 inflectional suffixes
 - $3873 \times 273 = 1,057,329$ total nouns
 - 658,410 after removing duplicate surface forms (case syncretism)

Surface Form	Lexical Form
mangteghaq	mangteghagh[N][ABS][SG]
mangteghaat	mangteghagh[N][ABS][PL]
mangteghaak	mangteghagh[N][ABS][DU]
mangteghaa	mangteghagh[N][ABS][SG][3SGPOSS]
:	

● EVALUATION OBJECTIVES

- Evaluate on a neutral dataset
 - Contrast performance with the FST analyzer
 - **NEUTRAL DATASET:** *Mrs. Della Waghiyi's St. Lawrence Island Yupik Texts With Grammatical Analysis* (Waghiyi & Nagai, 2001)
 - Identified **344 inflected nouns with no derivational morphology**
 - Supplemented the FST analyzer with a guesser module

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2020-2021

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- Results:

	Coverage (%)	Accuracy (%)
FST (No Guesser)	85.96	79.82
FST (w/Guesser)	100	84.50
Neural	100	91.81

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- An **out-of-vocabulary** (OOV) root is an unattested root that appears in the Waghiyi & Nagai (2001) evaluation dataset but does not appear in our data

OOV Root	FST	NN
aghnasinghagh	-	-
aghveghniigh	-	✓
akughvigagh	✓	✓
qikmiraagh	-	-
sakara	✓	-
sanaghte	-	-
tangiqagh	-	✓

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- A root with a **spelling variant** is one that differs in the Waghiyi & Nagai (2001) evaluation set from its form in our data

Root Variant	FST	NN
melqighagh	✓	✓
piitesiighagh	—	✓
uqfiiileghagh	—	✓
*ukusumun	—	✓

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Building a virtuous cycle

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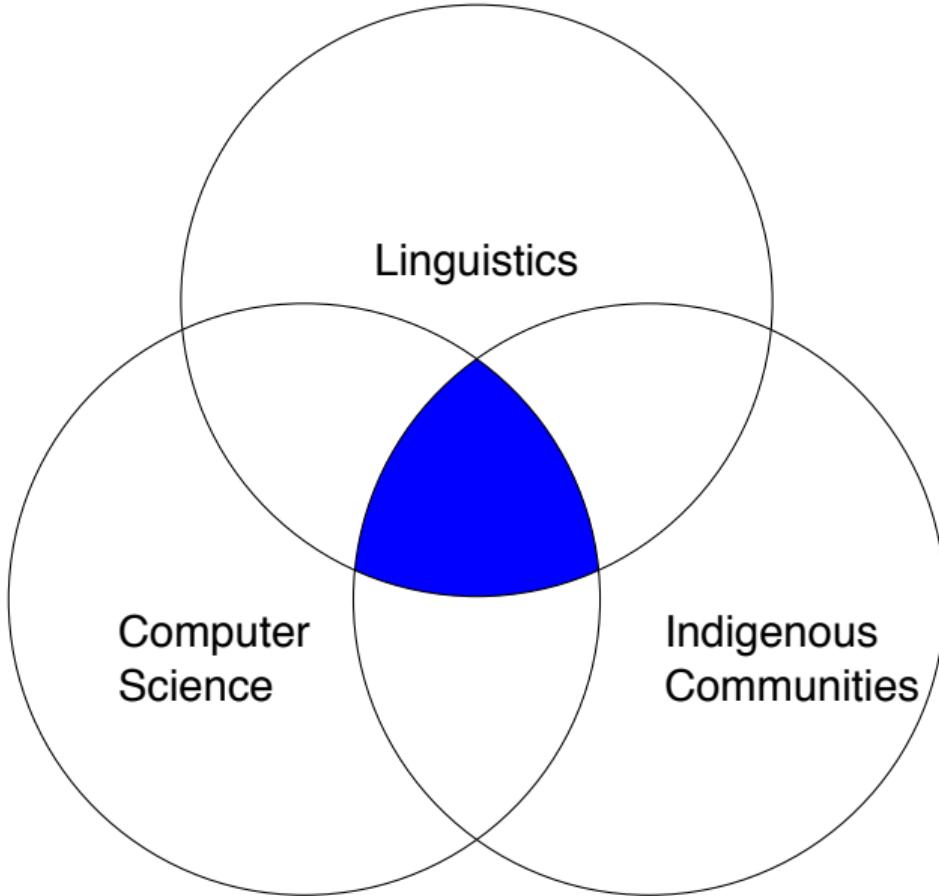
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- Digitization of legacy materials
 - Pedagogical materials & tools
 - Orthographic experimentation
 - Identify under-described phenomena
 - Real-time morphological analysis

Background

Intersecting machine learning & linguistic fieldwork

St. Lawrence Island Yupik

Intersecting machine learning & linguistic fieldwork

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Moving forward
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Moving forward

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$$p(\mathbf{e}) = p(e_t | e_1 \dots e_{t-1})$$

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$$\begin{aligned} p(\mathbf{e}) &= p(e_t | e_1 \dots e_{t-1}) \\ &\approx p(e_t | e_{t-1}) \end{aligned}$$

Background
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Moving forward
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Lane Schwartz

1. Zero-order approximation

XFOML RXKHRJFFJUJ ZLPWCFWKCYJ
FFJEYVKCQSGXYD QPAAMKBZAACIBZLHJQD

2. First-order approximation

OCRO HLI RGWR NMIELWIS EU LL NBNSEBYA TH EEI
ALHENHTTPA OOBTTVA NAH BRL

3. Second-order approximation

ON IE ANTSOUTINYS ARE T INCTORE ST BE S DEAMY
ACHIN D ILONASIVE TUOOWE AT TEASONARE FUSO
TIZIN ANDY TOBE SEACE CTISBE

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Lane Schwartz

5. First-Order Word Approximation

REPRESENTING AND SPEEDILY IS AN GOOD APT OR
COME CAN DIFFERENT NATURAL HERE HE THE A IN
CAME THE TO OF TO EXPERT GRAY COME TO FUR-
NISHES THE LINE MESSAGE HAD BE THESE.

6. Second-Order Word Approximation

THE HEAD AND IN FRONTAL ATTACK ON AN ENGLISH
WRITER THAT THE CHARACTER OF THIS POINT IS
THEREFORE ANOTHER METHOD FOR THE LETTERS
THAT THE TIME OF WHO EVER TOLD THE PROBLEM
FOR AN UNEXPECTED

$$\begin{aligned} p(\mathbf{e}) &= p(e_t | e_1 \dots e_{t-1}) \\ &\approx p(e_t | e_{t-1}) \end{aligned}$$

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$$p(\mathbf{e}) = p(e_t | e_1 \dots e_{t-1})$$

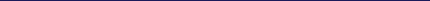
Background

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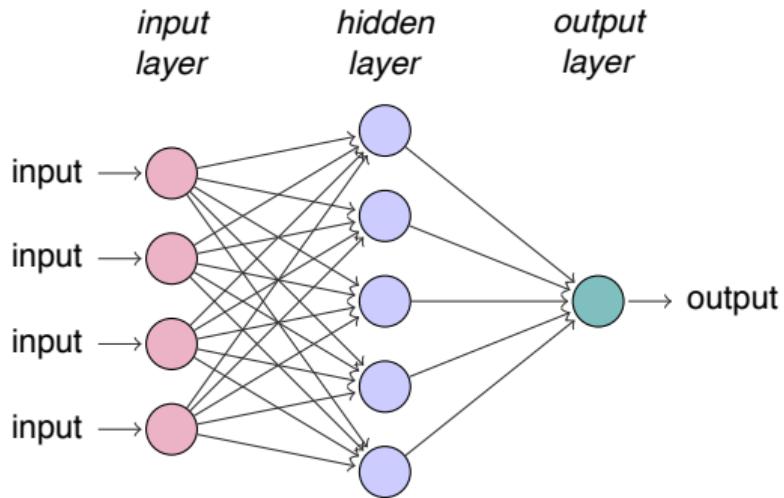


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Moving forward



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Moving forward

Intersecting machine learning & linguistic fieldwork

- Legacy text digitization
 - Web portal / interactive e-books
 - App-based dictionary
 - Language learning lessons
 - foma-based spell-checker
 - Forced aligner / speech recognizer
 - Machine translation

Background

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Moving forward
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Lane Schwartz

Feature-rich Open-vocabulary Interpretable Language Modelling

Interpretable Tensor Morpheme Representation

yaławtəma

“with the head”

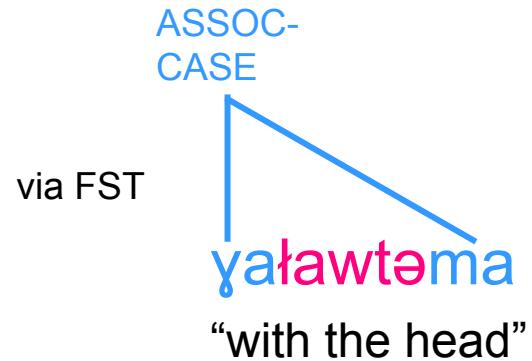
-/*ewət*- “head” (Chukchi)

Interpretable Tensor Morpheme Representation

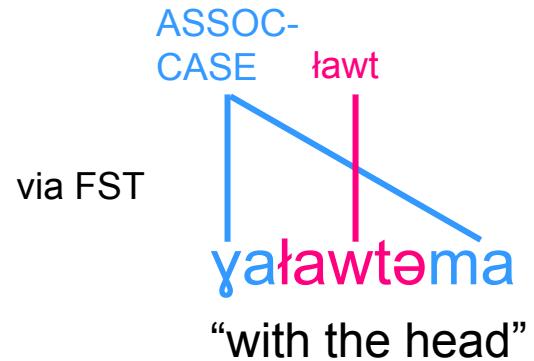
yaławtəma

“with the head”

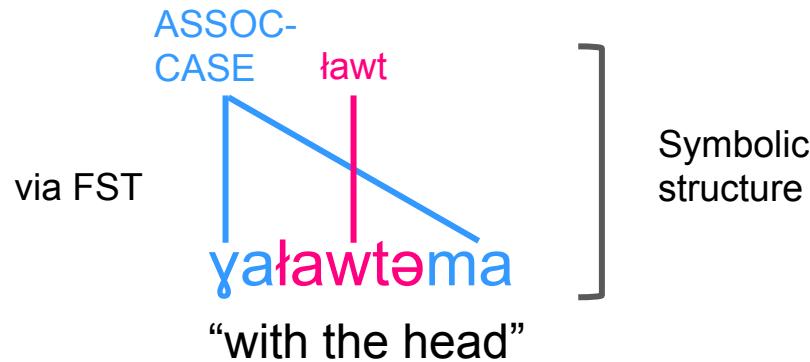
Interpretable Tensor Morpheme Representation



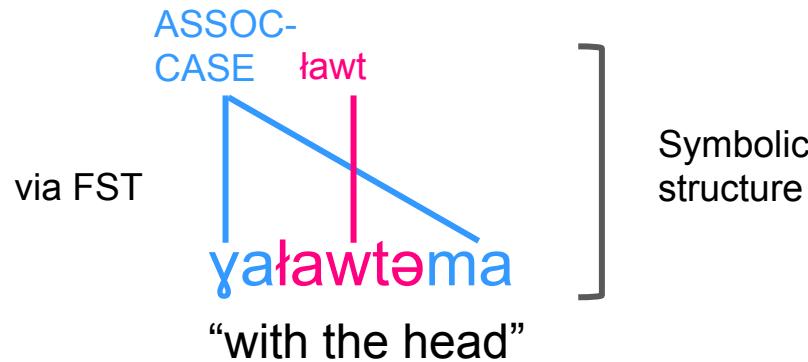
Interpretable Tensor Morpheme Representation



Interpretable Tensor Morpheme Representation

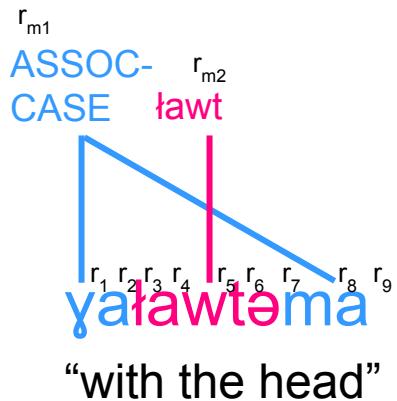


Interpretable Tensor Morpheme Representation



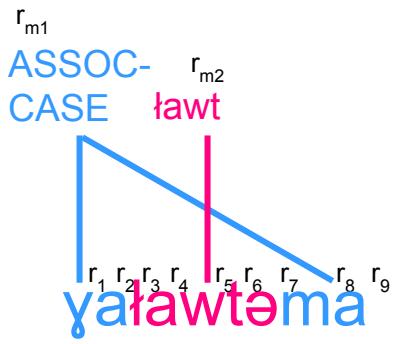
Symbolic structure → Tensor representation → Vector representation

Interpretable Tensor Morpheme Representation



Decompose into *fillers* and *roles*.
(Smolenksy 1990)

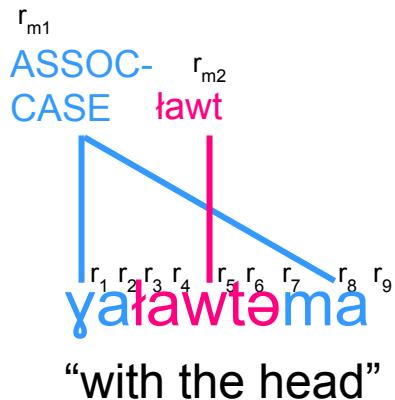
Interpretable Tensor Morpheme Representation



“with the head”

Embed the fillers and roles into *vectors*
(Smolenksy 1990)

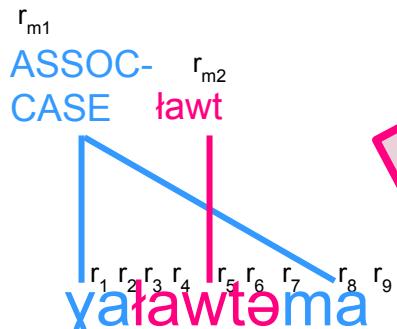
Interpretable Tensor Morpheme Representation



Embed the fillers and roles into *vectors*
(Smolenksy 1990)

$$Repr(ya-ma) = (\hat{y} \otimes \hat{r}_1 + \hat{a} \otimes \hat{r}_2 + \hat{m} \otimes \hat{r}_8 + \hat{a} \otimes \hat{r}_9) \otimes \hat{r}_{m_1}$$

Tensor Morpheme Representation



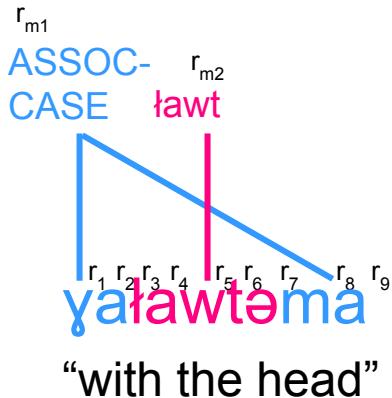
“with the head”

This means we have embeddings that are interpretable! ^_^

Embed the fillers and roles into *vectors*
(Smolenksy 1990)

$$Repr(ya-ma) = (\hat{y} \otimes \hat{r}_1 + \hat{a} \otimes \hat{r}_2 + \hat{m} \otimes \hat{r}_8 + \hat{a} \otimes \hat{r}_9) \otimes \hat{r}_{m1}$$

Interpretable Tensor Morpheme Representation



1. Deterministically create these with FST for known sequences
2. Learn them with neural model (e.g. RNN seq2seq) to generalize

Embed the fillers and roles into *vectors*
(Smolenksy 1990)

Deterministically construct morpheme tensors

- a. Run morphological analyzer on training data to identify morphemes

qikmighhaak → ^a qikmigh - ghagh - [Abs.Du]
“Two small dogs” “dog - small.N - [Abs.Du]”

Deterministically construct morpheme tensors

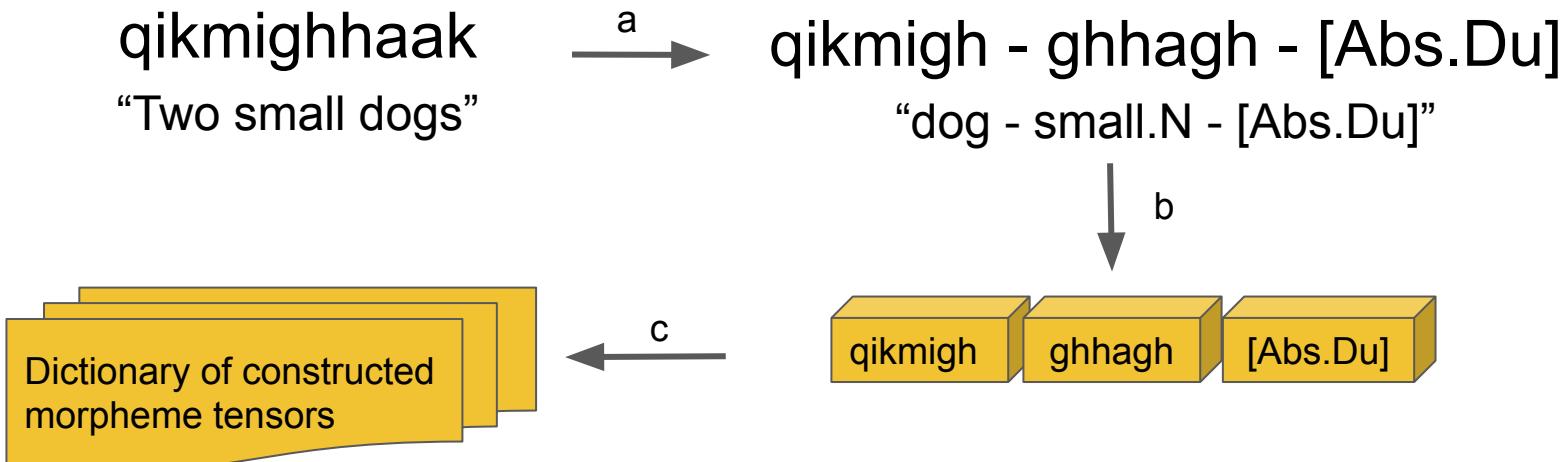
- a. Run morphological analyzer on training data to identify morphemes
- b. Use **Tensor Product Representation** to deterministically calculate morpheme tensors



(St. Lawrence Island Yupik)

Deterministically construct morpheme tensors

- a. Run morphological analyzer on training data to identify morphemes
- b. Use **Tensor Product Representation** to deterministically calculate **morpheme tensors**
- c. Save these morpheme tensors for later use as gold standard labels



Autoencoder

Dictionary of constructed
morpheme tensors

High dimensionality:
 $10^3 - 10^9$ floats per vector

qikmigh

ghhagh

[Abs.Du]

Dictionary of constructed
morpheme tensors

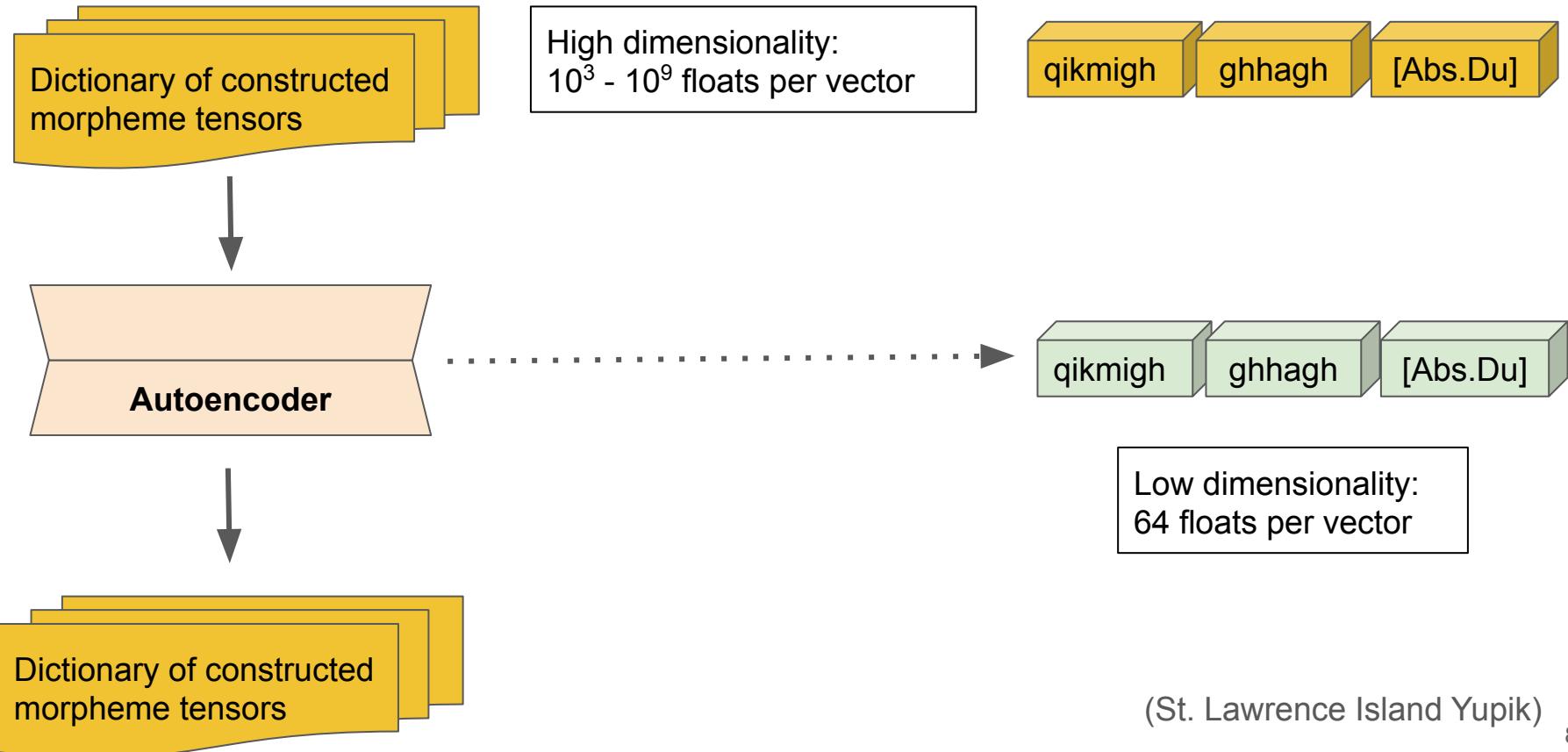
High dimensionality:
 $10^3 - 10^9$ floats per vector

qikmigh ghhagh [Abs.Du]

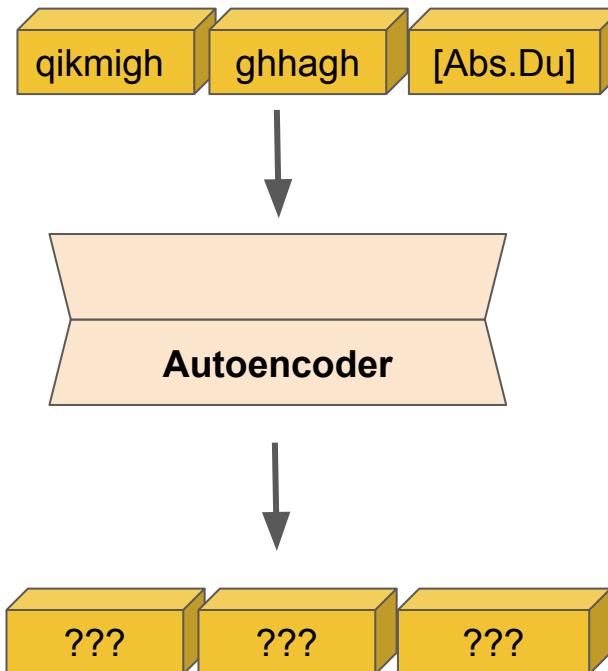
qikmigh ghhagh [Abs.Du]

Low dimensionality:
64 floats per vector

Use autoencoder to learn morpheme vectors



Problem: Morpheme tensors are sparse



As a result, learning signal is very weak.

Solution: Unbinding Loss

