

# Incorporating Sociolinguistic Theories for a Better Modeling of the Arabic Varieties

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# Arabic 101

- Spoken in 22 Arab Countries.
  - + Minority groups in other non-Arab countries.



# Arabic 101

- Spoken in 22 Arab Countries.
  - + Minority groups in other non-Arab countries.
- Vast geographical area ⇒ Variation.
- Additionally, impact of:
  - Other local languages in the region  
(e.g., Tamazight, Coptic)
  - Colonial languages  
(e.g., English, French, Italian)
  - Contact languages  
(e.g., Greek, Persian)



# Example of Lexical Variation



Source: <https://www.youtube.com/shorts/WXL4xITDreI>

# Example of Lexical Variation



# Non-negligible Degrees of Variation in:

- ① Phonology
- ② Morphology
- ③ Lexicon
- ④ Syntax (e.g., Word Order)
- ⑤ Semantics (e.g., False Friends)
- ⑥ Culture?



# Non-negligible Degrees of Variation in:

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- ④ Syntax (e.g., Word Order)
- ⑤ Semantics (e.g., False Friends)
- ⑥ Culture?

Interspeaker Variation !

# Theory of Intraspeaker Variation #1: Diglossia (Ferguson, 1959)

# Diglossia

“a language state in which two varieties of the language co-exist within the same speaking community:

- a high variety linked to higher prestige
  - Modern Standard Arabic (MSA)
- a low variety perceived to be of lower status.
  - Arabic Dialects”

- MSA was the language of literary
  - e.g., books, newspapers, ...
- Arabic Dialects increasingly written online
  - e.g., texting, social media, ...

# Operationalization of *Diglossia* in NLP

- Dialect Identification (i.e., sentence -> dialect)

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- MSA as an independent dialect.

# Two Limitations

## ① Disjointedness

- A sentence valid in a dialect **can not** be valid in another dialect.

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## ② **Binarization**

- Dialectal sentences are **equally divergent from MSA**.



# The Binarization Limitation

# Different ways of saying: happy 😊



فرج  
farih

مبسوط  
mbsoT

مشهیص  
mʃhys

Level of  
Dialectness



# Different ways of saying: happy 😊

**MSA root meaning** Root

to be happy



فرح  
frh

فَرِحَ  
farih

مبسوط  
mbsoT

مشهِيص  
mʃhys

Level of  
Dialectness



# Different ways of saying: happy 😊

**MSA root meaning**

to be happy

**Root**



فَرَحٌ  
frh



فَرِحٌ  
farih

extend - cheer

بَسْطٌ  
bsT

مَبْسُوتٌ  
mbsoT

N/A

شَيْحُصٌ  
ʃys

مَشَيْحُصٌ  
mʃys

**Level of  
Dialectness**



# Theory of Intraspeaker Variation #2: Dialect Levels (Badawi, 1973)

# Dialect Levels

- focused on spoken language
- identified **five** different dialect levels
- with examples of linguistic features for each level

# Dialect Levels

**Note:** Fus-ha **فصحي** is the term Arabs use for the standardized classical and modern varieties.

- ① Heritage Fus-ha
  - ② Fus-ha of the age (we live in)
  - ③ Dialect of the (well-)educated
  - ④ Dialect of the Literate
  - ⑤ Dialect of the Illiterate
- ↓
- Level of  
Dialectness**

# Operationalization of *Dialect Levels* in NLP

Few efforts mainly proposing guidelines and annotating limited data:

① (Habash, 2008), (Elfardy, 2012)

- token-level annotations mapped to sentence-level
- data mostly MSA
- not publicly available

② (Zaidan, 2011)

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Habash, Nizar et al. "Guidelines for annotation of Arabic dialectness." Workshop on HLT & NLP - LREC 2008.

Elfardy, Heba and Diab, Mona. "Simplified guidelines for the creation of Large Scale Dialectal Arabic Annotations." LREC 2012.

Zaidan, Omar F. and Callison-Burch, Chris. "The Arabic Online Commentary Dataset: an Annotated Dataset of Informal Arabic with High Dialectal Content."

# AOC Dataset



## Arabic Online Commentary Dataset (Zaidan et. al, 2011)

- Comments to news articles
- Three publishers (Egypt, Jordan, Saudi Arabia)
- 127,835 sentences (3 annotations each)
- Popular Dialect Identification (DI) labels

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Zaidan, Omar F. and Callison-Burch, Chris. "The Arabic Online Commentary Dataset: an Annotated Dataset of Informal Arabic with High Dialectal Content."

# AOC Dataset



## Arabic Online Commentary Dataset (Zaidan et. al, 2011)

- Comments to news articles
- Three publishers (Egypt, Jordan, Saudi Arabia)
- 127,835 sentences (3 annotations each)
- Popular Dialect Identification (DI) labels
- Ignored *Discrete Level of Dialectness* labels!

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Zaidan, Omar F. and Callison-Burch, Chris. "The Arabic Online Commentary Dataset: an Annotated Dataset of Informal Arabic with High Dialectal Content."

Tell us how much dialect (عامية) is in the sentence.



Tell us how much dialect (عامية) is in the sentence.



Fleiss'  $\kappa = 0.44$

Tell us how much dialect (عامية) is in the sentence.



Fleiss'  $\kappa = 0.44$



Embrace annotators disagreement!

# Sentence with two valid pronunciations

نبدأ بـالشغل الصحيح في تطوير المدارس وتوفير

المراقبين عليها

We start with the right task of developing schools and providing observers over them

# Sentence with two valid pronunciations

نبتدى بقى الشغل الصح في تطوير المدارس وتوفير

المراقبين عليها

We start with the right task of developing schools and providing observers over them



Little



Little



Most



## From AOC to AOC-ALDi

- 1 Labels into numeric values
- 2 Algebraic Mean

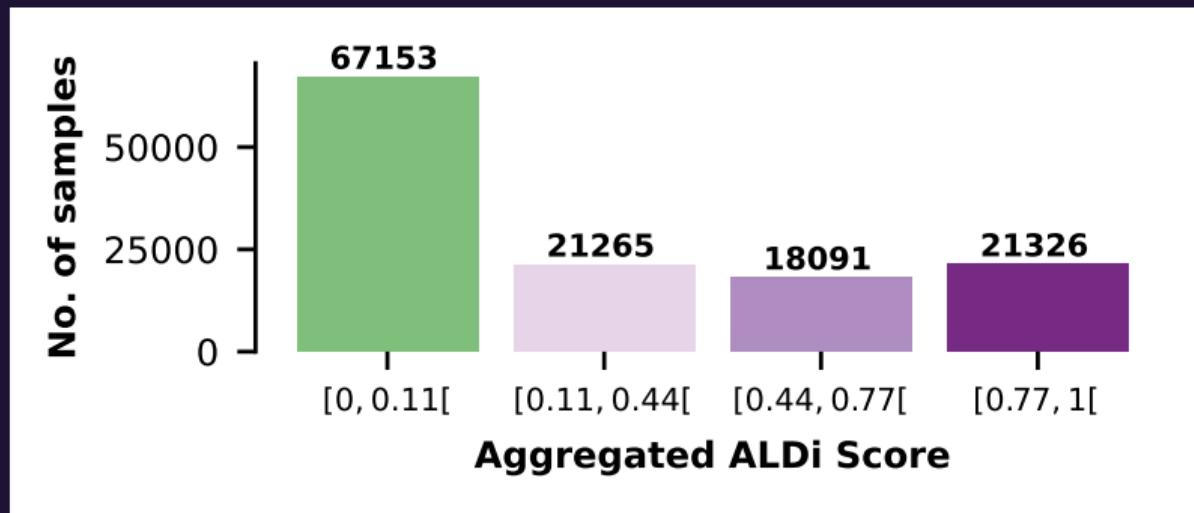


e.g., ALDi(MSA, MSA, Little) =  $\overline{(0, 0, \frac{1}{3})} = \frac{1}{9} \approx 0.11$



Krippendorff's  $\alpha$  (interval) = 0.63

# ALDi scores Distribution



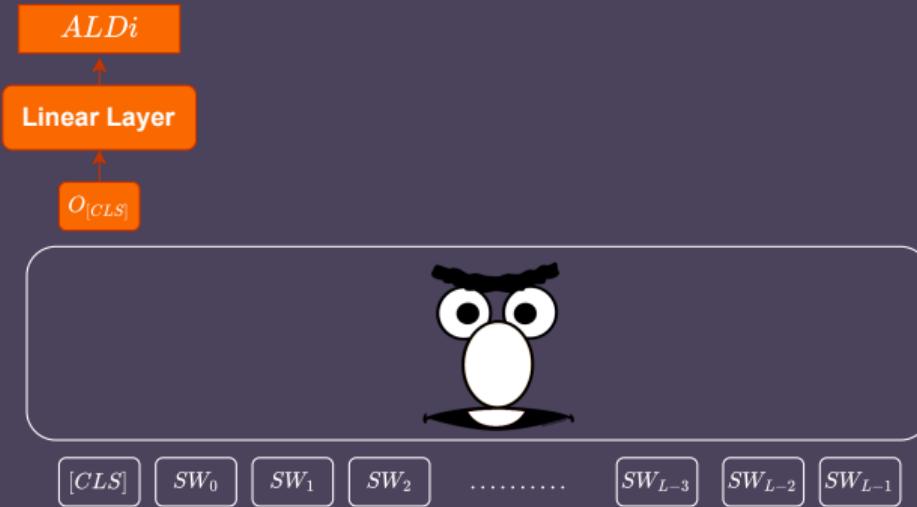
# ★★★ Our Operationalization ★★★

- **Arabic Level of Dialectness (ALDi):**  
Divergence from Standard Arabic (MSA).
- Continuous score in [0, 1].
- Sentence-like level.

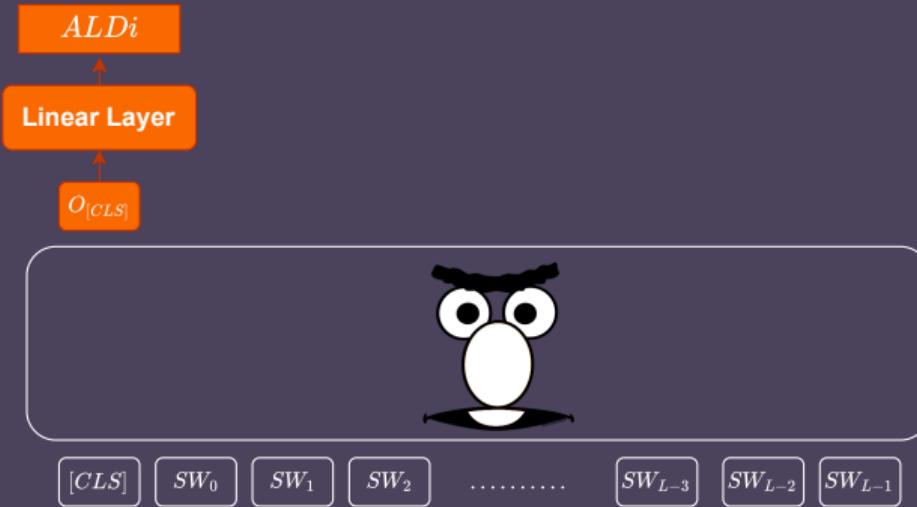


# Building a model to estimate ALDi automatically

# Sentence-ALDi model

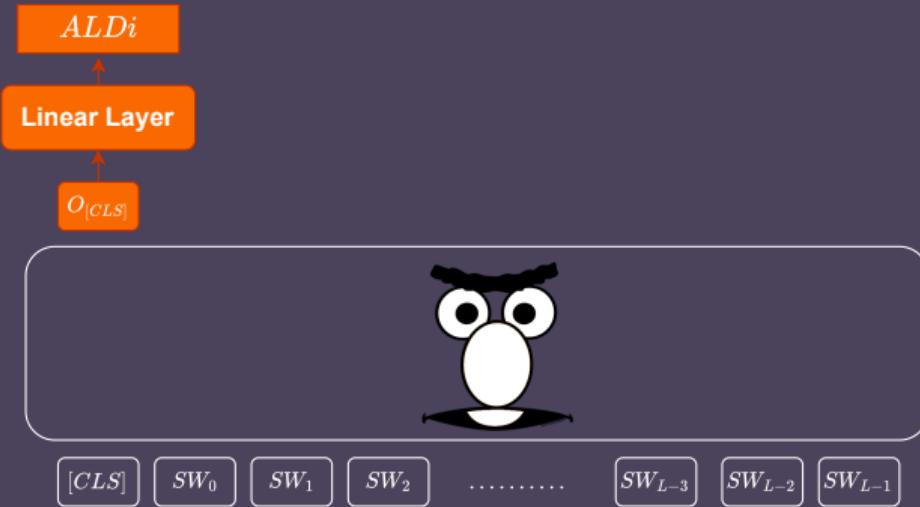


# Sentence-ALDi model



Dialect-agnostic

# Sentence-ALDi model



Dialect-agnostic



RMSE(test set) = 0.18



# Applications of ALDi

## ① Analyzing Intraspeaker Variation (Presidential Speeches)

# Case Study - Presidential speeches

- Arab presidents use:
  - Modern Standard Arabic (MSA) - authority
  - Dialectal Arabic (DA) - compassion and belonging
- Replicated for speeches of former Tunisian and Egyptian presidents.

# Egyptian President (El-Sisi) - 18/07/2022



Source: <https://beta.sis.gov.eg/en/presidency/presidents-monthly-activities/presidents-activities-during-july-2022/>

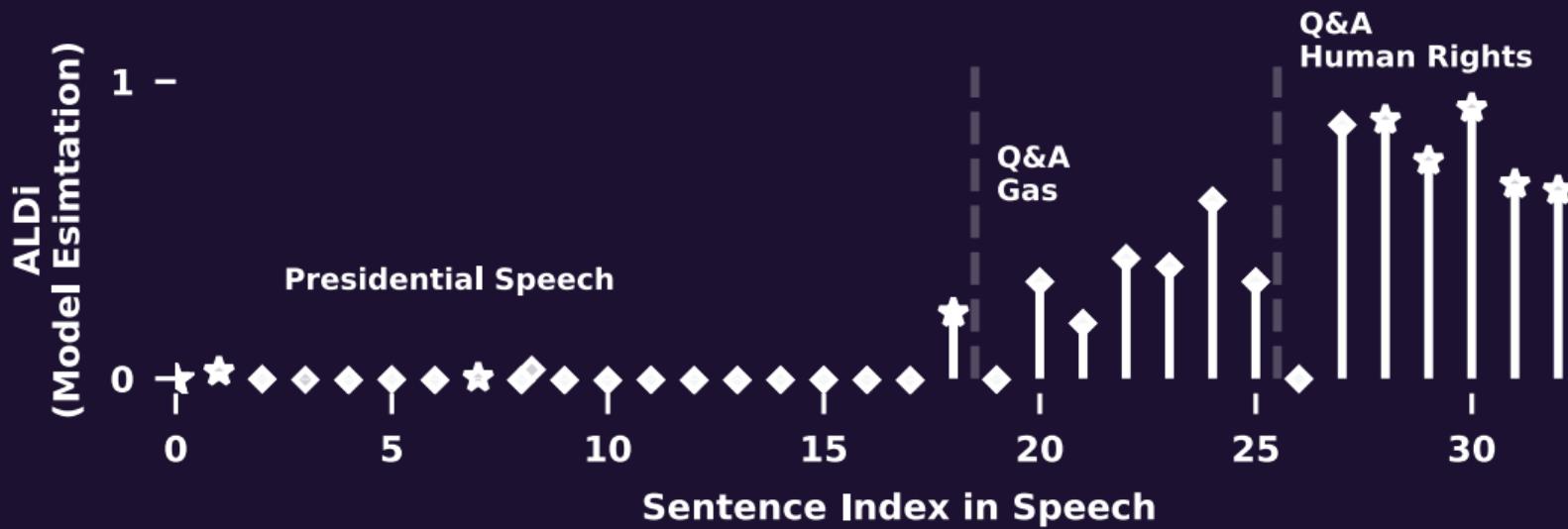
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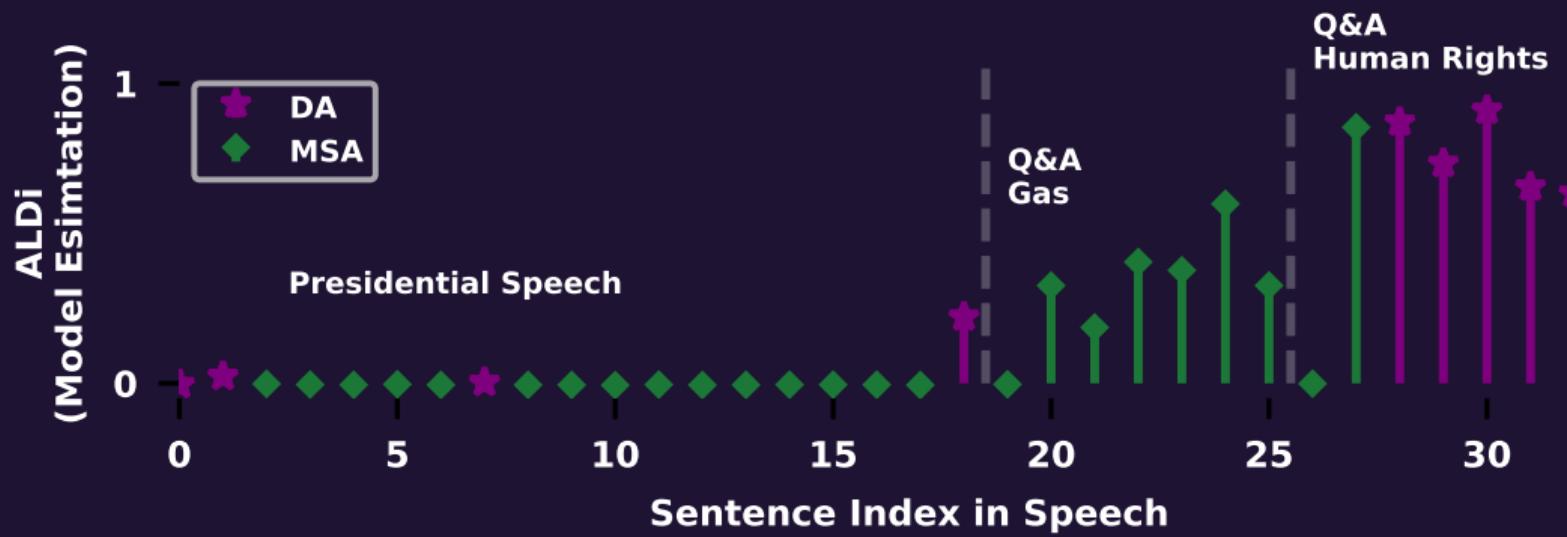
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# Applications of ALDi

## ② Impact of Interannotator Agreement

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Keleg, Amr, Magdy, Walid, and Goldwater, Sharon. "Estimating the Level of Dialectness Predicts Inter-annotator Agreement in Multi-dialect Arabic Datasets." ACL 2024.

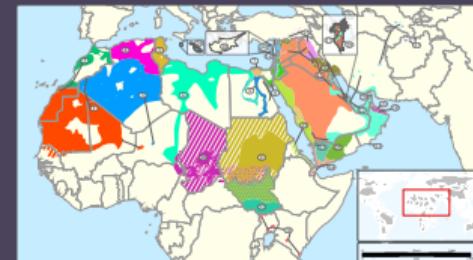
## 2) Annotating Multi-Dialect Arabic Datasets

**Common Practice:** 🎲 randomly assign to Arabic speakers

☢️☢️☢️ **Annotator's dialect ≠ Sample's dialect** ☢️☢️☢️

📎 More strict annotating Hate Speech 🚫  
(Bergman and Diab, 2022)

📎 Less accurate identifying Sarcasm 😛  
(Abu Farha and Magdy, 2022)



Bergman, A. and Diab, Mona. "Towards Responsible Natural Language Annotation for the Varieties of Arabic."

Abu Farha, Ibrahim and Magdy, Walid. "The Effect of Arabic Dialect Familiarity on Data Annotation."

# Annotation Codebook (v1.0)



**Step 1:** Identify the dialect of each sample

**Step 2:** Route the sample to speakers of its dialect

# Annotation Codebook (v1.0)



🏷 Step 1: Identify the dialect of each sample

✖ Step 2: Route the sample to speakers of its dialect

**Hard to crowdsource speakers of some dialects  
(i.e., Limited resource)** 💎💎

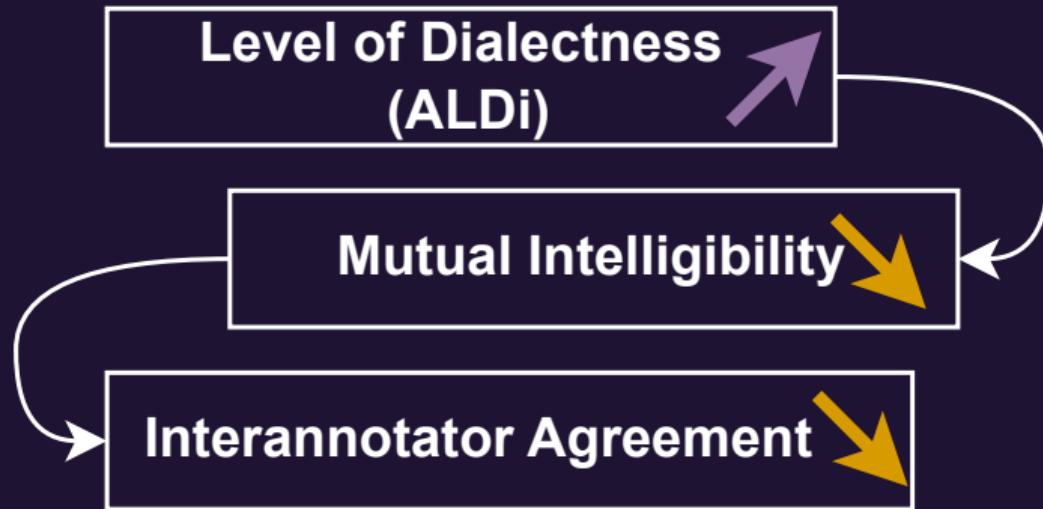
(Mubarak and Darwish, 2016)



# Should some dialectal samples be prioritized?



# Intuition





# Analysis

15 public datasets covering 6 Tasks:  
Hatespeech, Sentiment Analysis, Dialect Identification, ...

- (1) sentence-level classification datasets**
- (2) multi-dialect samples**
- (3) samples randomly assigned to annotators**
- (4) individual annotator labels**



## Methodology:

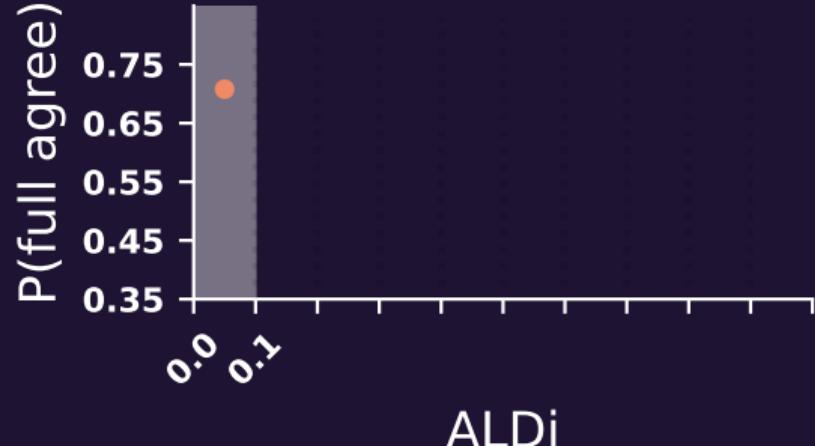
- ① Estimate ALDi of samples.
- ② Bin samples.
- ③  $P_{\text{bin}}(\text{Full Agreement})$

$$P_{\text{bin}}(\text{Full Agreement}) \approx \frac{N_{(\text{bin})} \text{ Full Agreement}}{N_{(\text{bin})} \text{ Total Samples}}$$



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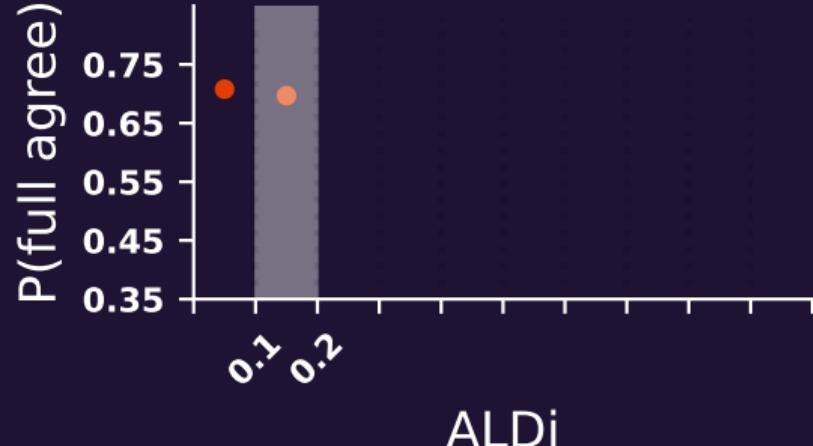


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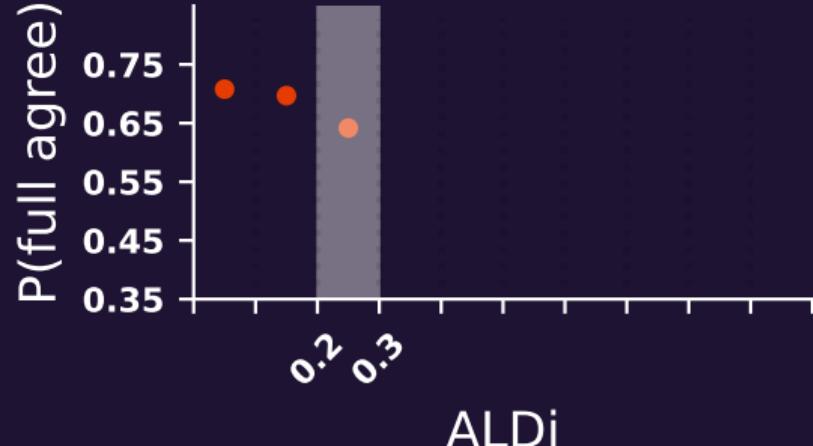


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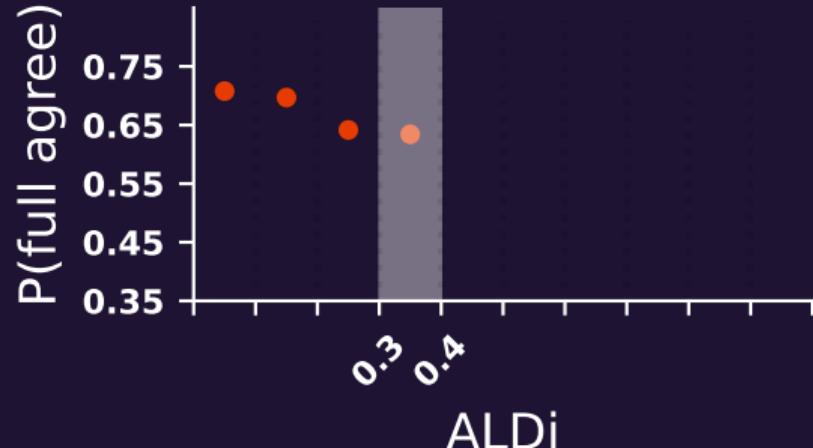


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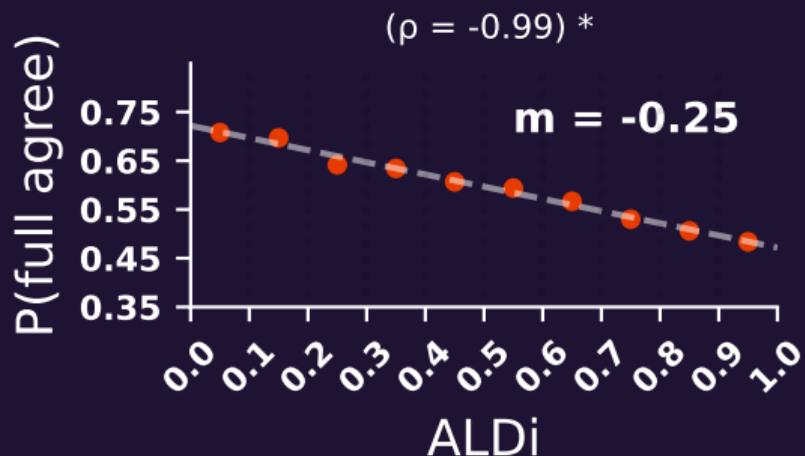


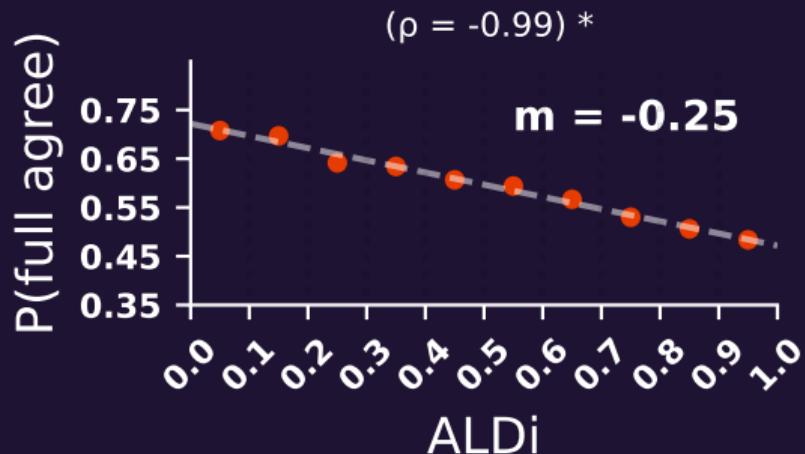
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Finding (1) - For 8 of 12 non Dialect Identification datasets

ALDi ➔

Interannotator Agreement ➔

with significant strong negative  $\rho < -0.7$

# IAA - Dialect Identification Dataset

 **Labels (Macro-regional):**

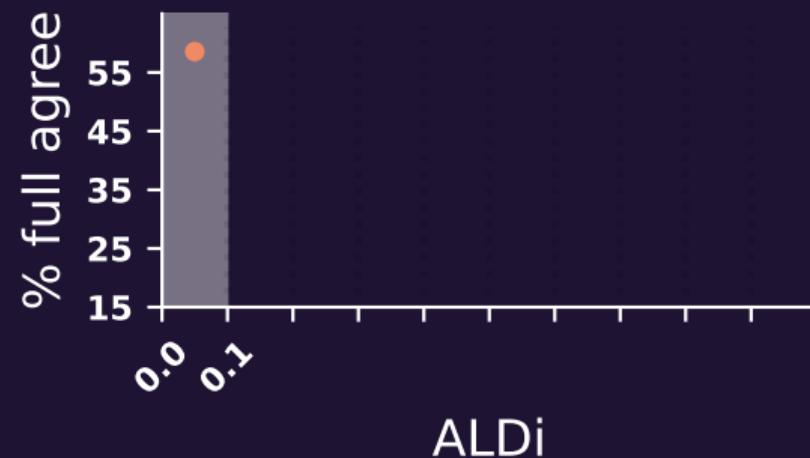
MSA, Maghreb, Egypt, Levant, Gulf



# IAA - Dialect Identification Dataset

## i Labels (Macro-regional):

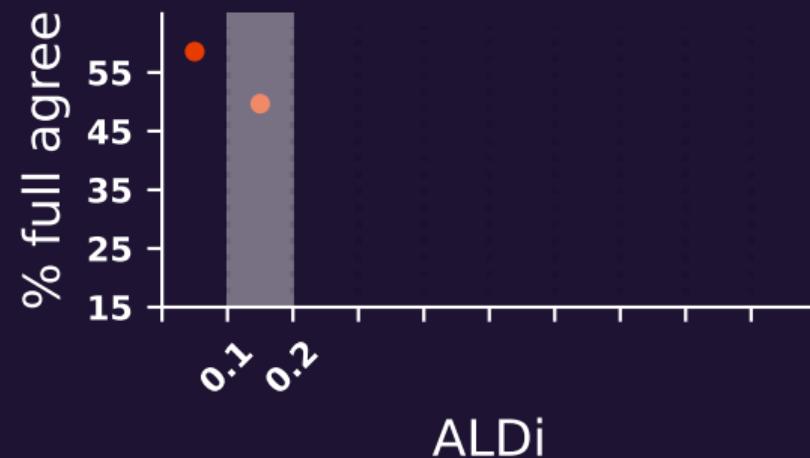
MSA, Egypt, Gulf, Levant, Maghreb



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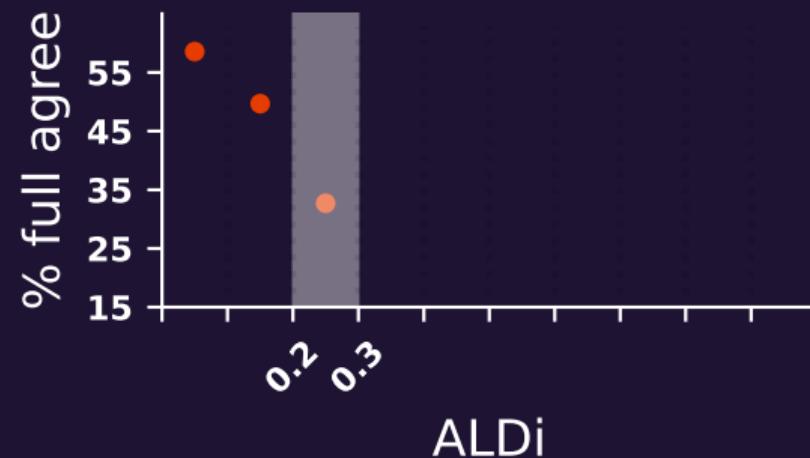
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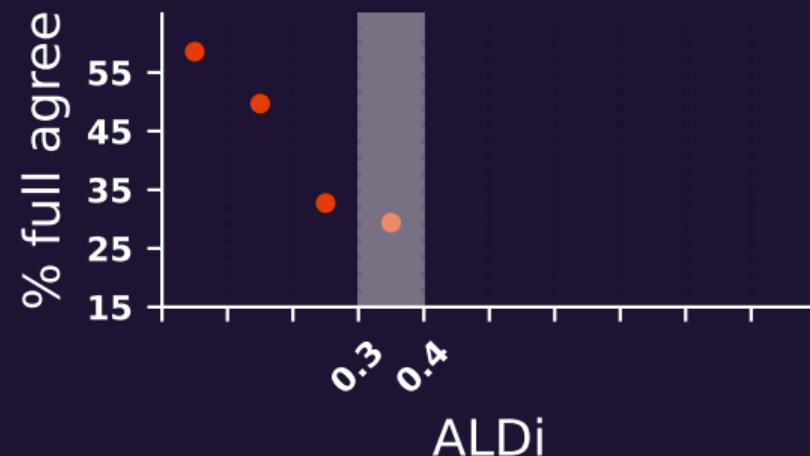
MSA, Egypt, Gulf, Levant, Maghreb



# IAA - Dialect Identification Dataset

## i Labels (Macro-regional):

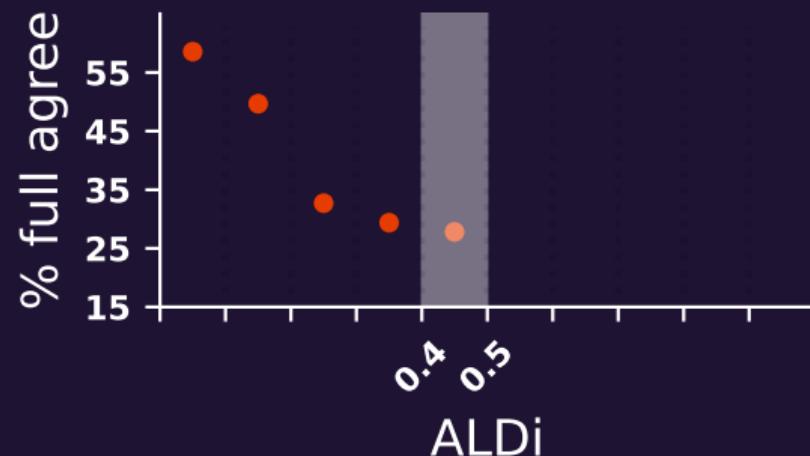
MSA, Egypt, Gulf, Levant, Maghreb



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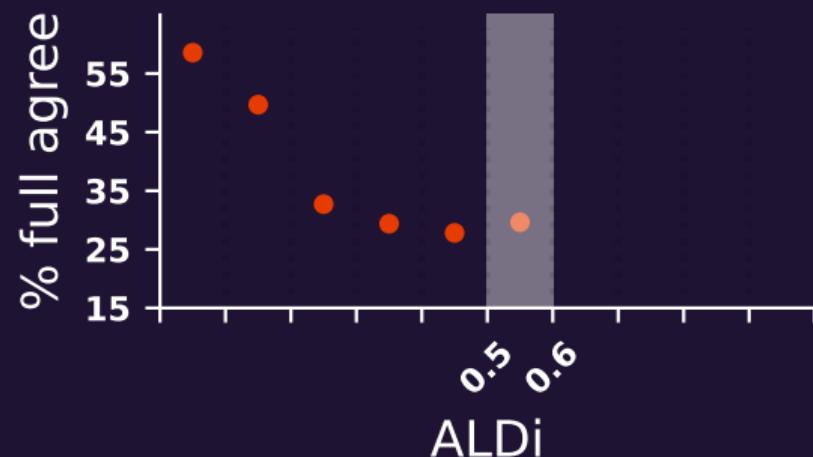
MSA, Egypt, Gulf, Levant, Maghreb



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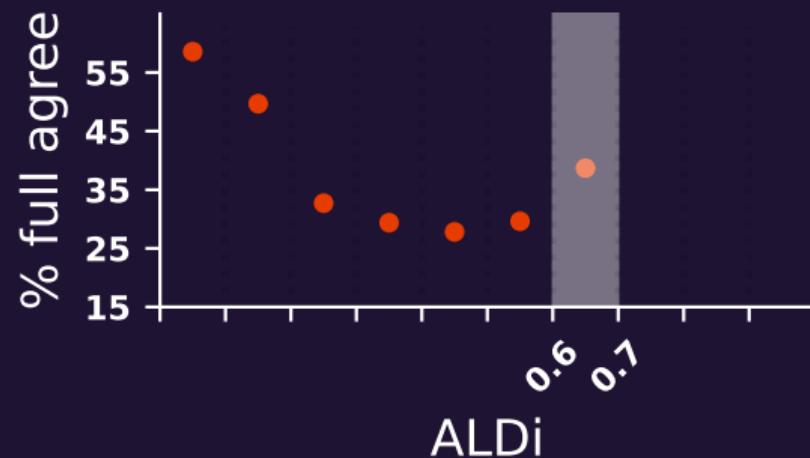
MSA, Egypt, Gulf, Levant, Maghreb



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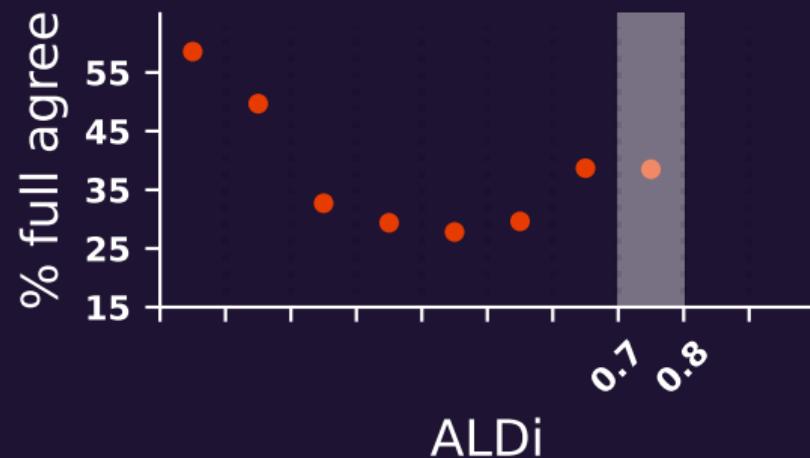
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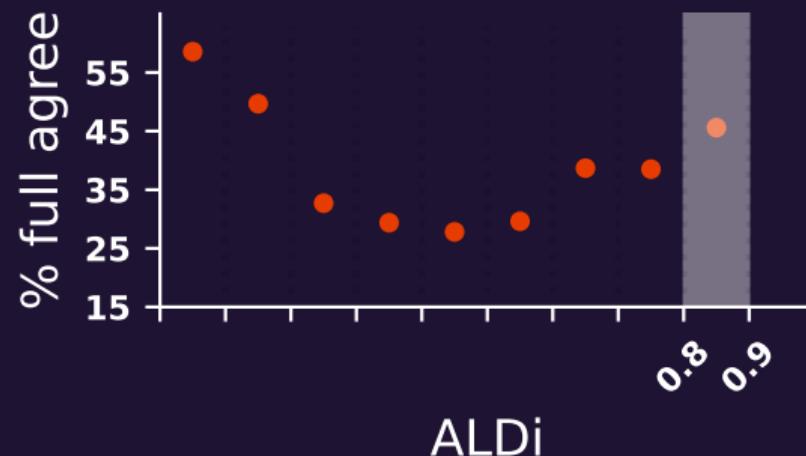
MSA, Egypt, Gulf, Levant, Maghreb



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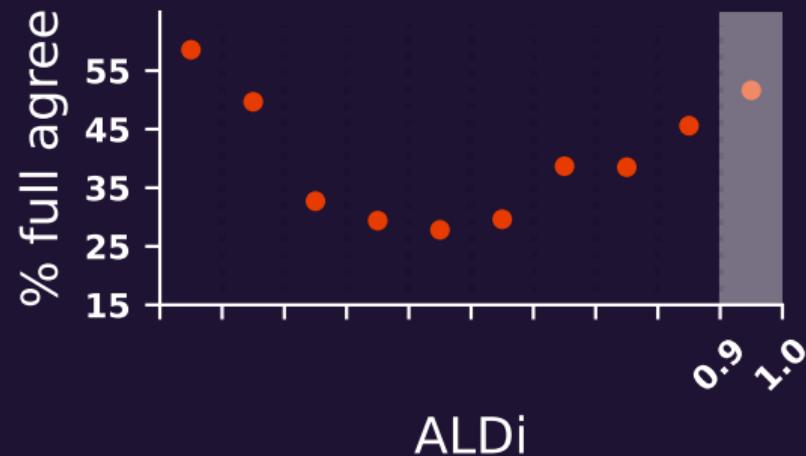
MSA, Egypt, Gulf, Levant, Maghreb

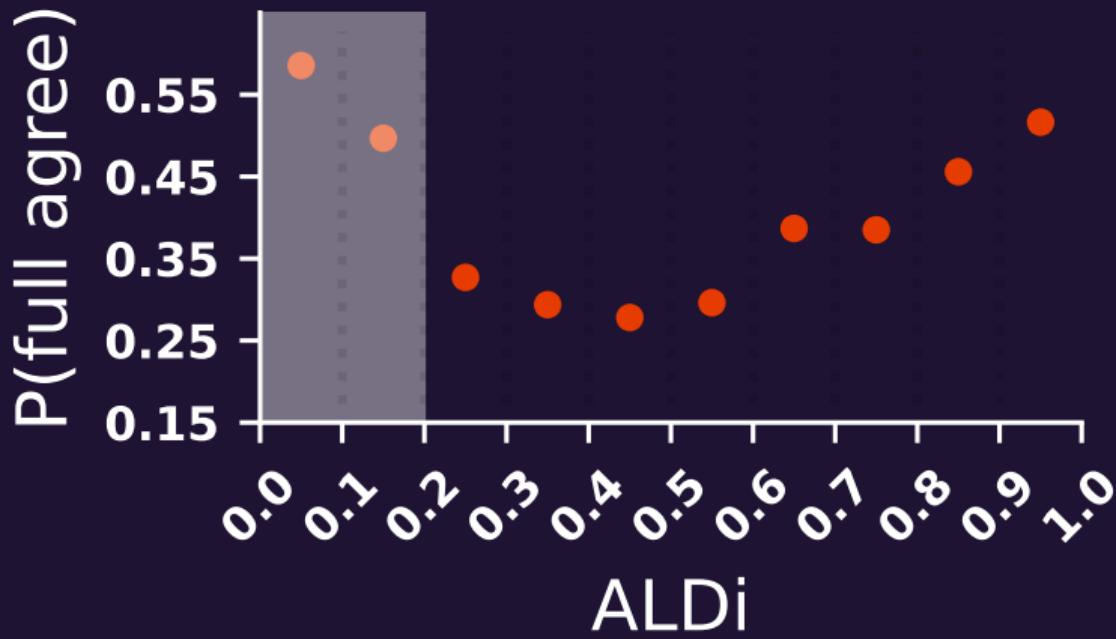


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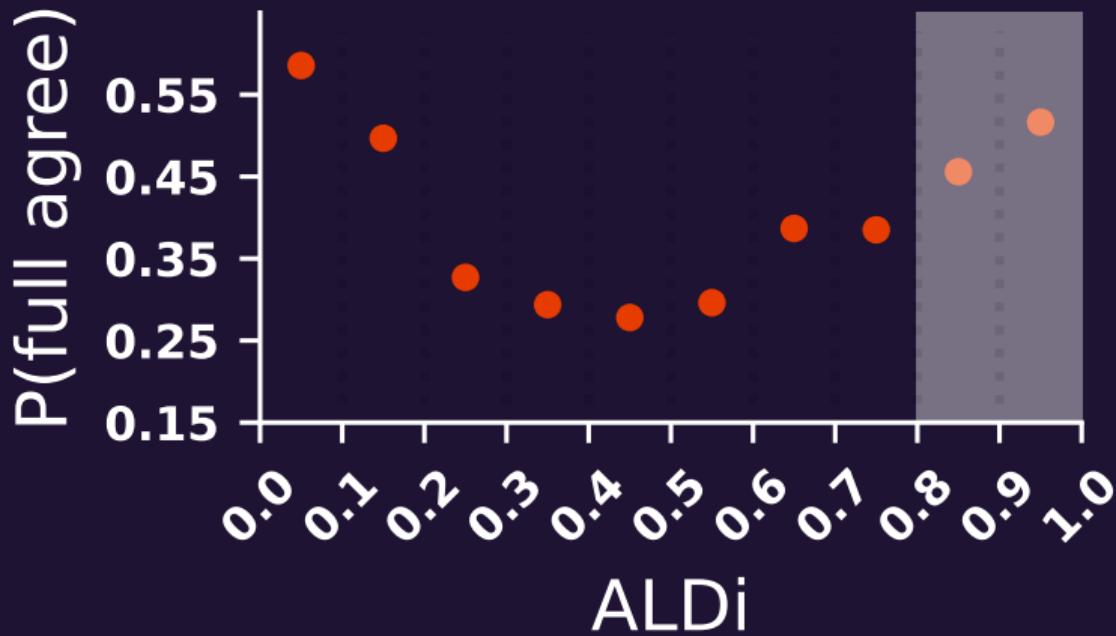
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MSA, Egypt, Gulf, Levant, Maghreb

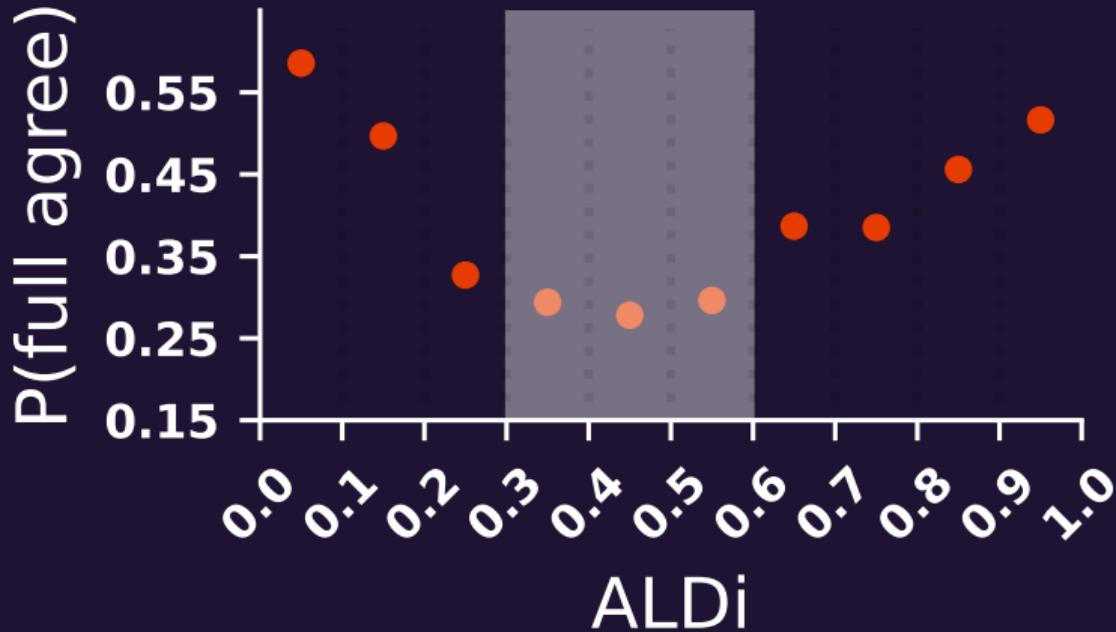




📎 MSA samples



Highly DA samples (with distinctive cues?)



- 📎 1) Hard to determine the dialect?
- OR
- 📎 2) Valid in multiple dialects?

# Annotation Codebook (v1.1)



- 💡 Prioritize routing high-ALDi samples to speakers of the samples' respective dialects, (Finding 1)
- 🔍 for which Dialect Identification is more accurate. (Finding 2)

# Thanks!



@amr-keleg.bsky.social

# Thanks!



@amr-keleg.bsky.social

## Summary

- 1 Arabic sentences exist along a continuum
  - Pure MSA < ----- > Highly Dialectal
- 2 Adapting sociolinguistic theories can improve our NLP tools



Please enlighten me about variation in your native languages!

# References I

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Abu Dhabi, United Arab Emirates (Hybrid): Association for Computational Linguistics, pp. 399–408. DOI:  
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URL: <https://aclanthology.org/P11-2007>.

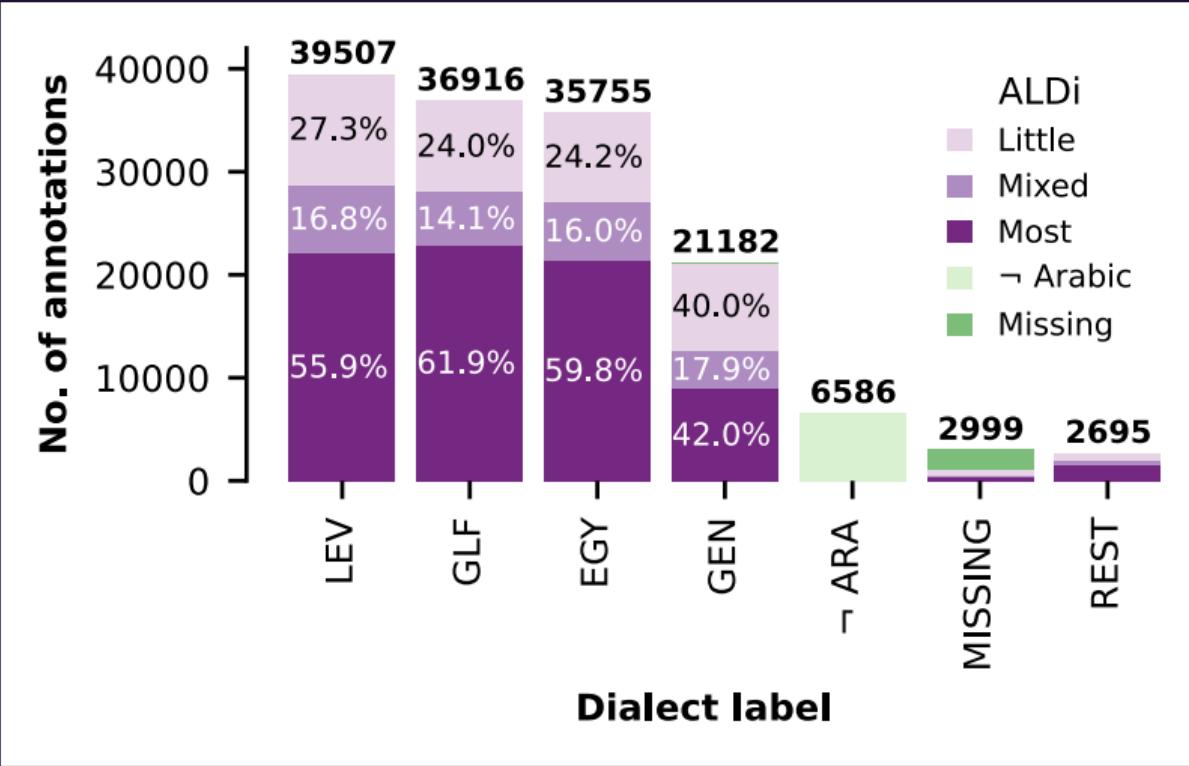
# Exposure is key for Intelligibility

"Egyptian Arabic and to a lesser extent Levantine Arabic are **widely understood** because of the **massive exposure** to them through the media and the arts during the last generation or so." (S'hiri, 2002)

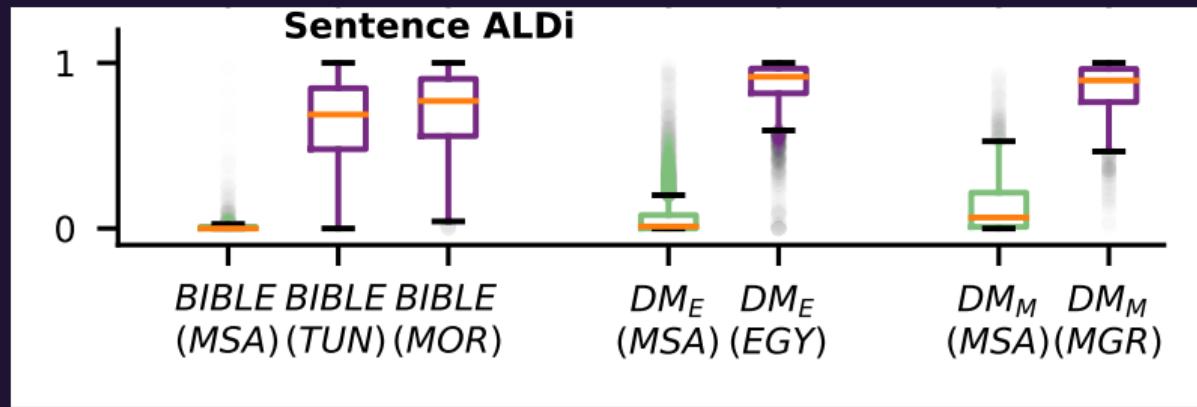
Sometimes compared to Romance Languages.

# Why not different languages?

"MSA is a kind of **communally-owned reservoir** that Arabs use to ORANGE make themselves understood to others from distant countries". (Holes, 1995)



ALDi scores automatically estimated:



# NADI 2024 Dataset

Is it possible that the tweet is authored by someone who speaks one of your country's dialects?

- 1,120 sentences.
- with geolocations uniformly distributed across 14 countries.
- 3 annotators from 9 different countries (total of 27)

# NADI 2024 Dataset

Is it possible that the tweet is authored by someone who speaks one of your country's dialects?

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Sentence	Valid in
وين يلعب هذا ما شفته	Algeria  , Palestine  , Yemen 

---

IFF an annotator labels the tweet as written in one of their country-level dialects.

Please evaluate the Level of Dialectness of each tweet as:

**L0** Sound MSA

**L1** Formal Colloquial or Colloquial-influenced MSA

**L2** Natural/Ordinary Colloquial

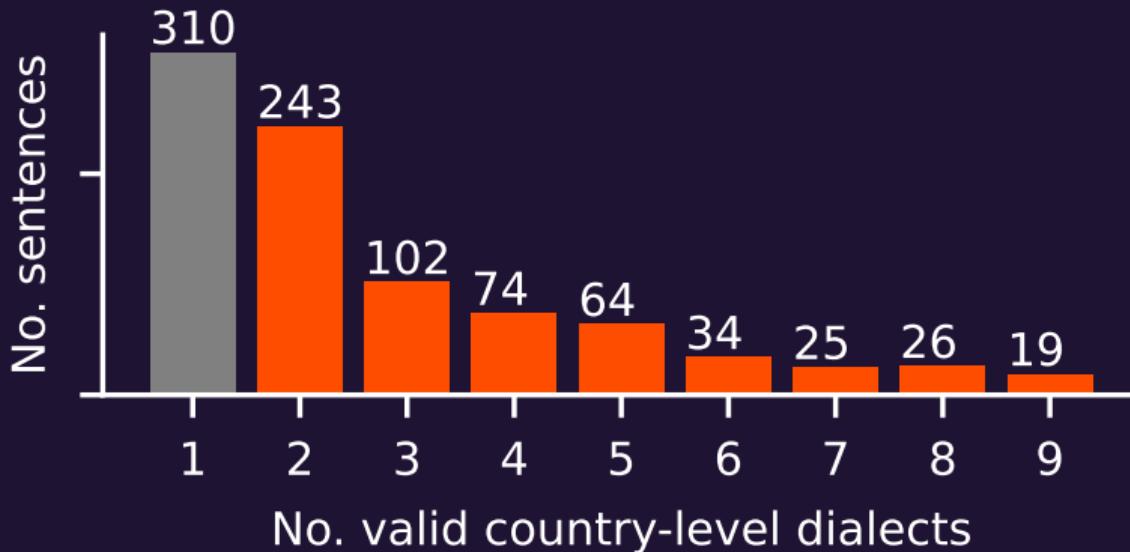
**L3** Informal (or Vulgar) Colloquial

Country	N valid	Krip. $\alpha$
Algeria	333	0.66
Morocco	230	0.74
Tunisia	189	0.75
Egypt	353	0.82
Sudan	393	0.66
Palestine	375	0.68
Syria	475	0.79
Iraq	271	0.73
Yemen	454	0.50



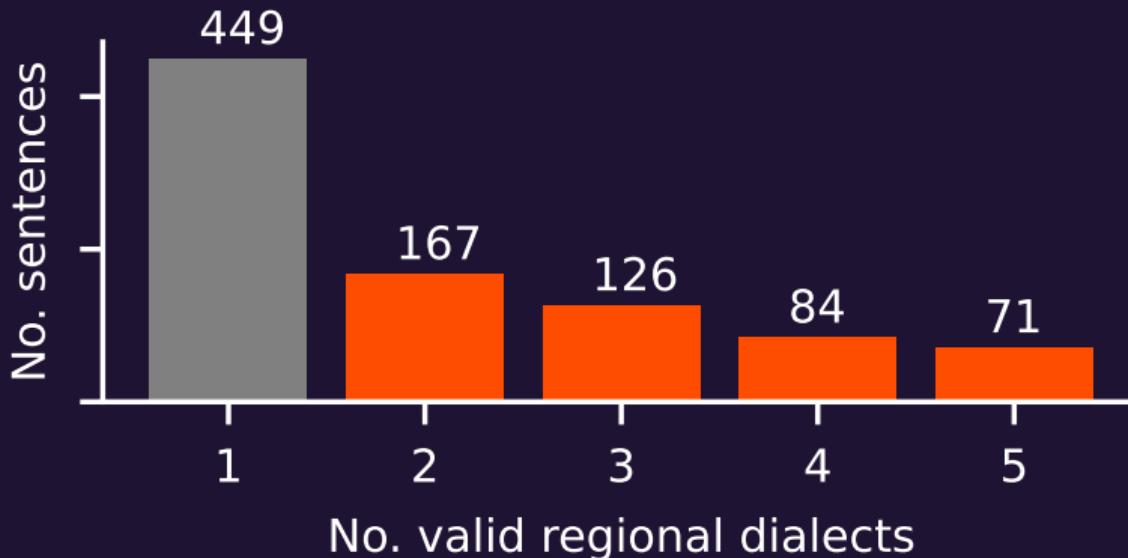
Improved alpha scores than AOC-ALDi.

# Multilabel samples in NADI 2024?



All samples but 310 are multi-dialect (country level).

# Multilabel samples in NADI 2024?



- 💡 > 50% of samples are valid in multiple regions.
- ✖️ Not just because of within-region similarities!