

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 \Rightarrow 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?



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- 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 😊



فَرِحَ
farih

مبسوط
mbsoT

مشييص
mjfhys

Level of
Dialecttness



Different ways of saying: happy 😊

MSA root meaning

to be happy

Root

فرح
frh



فرح
farih

مبسوط
mbsoT

مشيخص
mjfhys

**Level of
Dialecttness**



Different ways of saying: happy 😊

MSA root meaning

to be happy

Root



فرح

frh

فرح

farih

extend - cheer

بسط

bsT

مبسوط

mbsot

N/A

شهيص

fhys

مشهيص

mjhys

**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 **فصحى** 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
Dialecttness**



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)

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

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.

Dialect Level	كمية اللهجة العامية
✓ Choose level...	
No dialect (فصحى فقط)	
A bit of dialect (القليل من العامية)	
Mixed (خليط من الفصحى والعامية)	
Mostly dialect (معظمها عامية)	
Not Arabic (لغة أخرى أو رموز)	
Choose level...	▼
Choose level...	▼

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Fleiss' $\kappa = 0.44$

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Fleiss' $\kappa = 0.44$



Embrace annotators disagreement!

Sentence with two valid pronunciations

نبتدى بقى الشغل الصح فى تطوير المدارس وتوفير
المراقبين عليها

We start with the right task of de-
veloping schools and providing ob-
servers over them

Sentence with two valid pronunciations

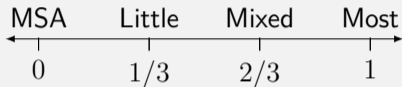
نبتدى بقي الشغل الصح فى تطوير المدارس وتوفير
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We start with the right task of developing schools and providing observers over them

  
Little Little Most

 From AOC to AOC-ALDi 

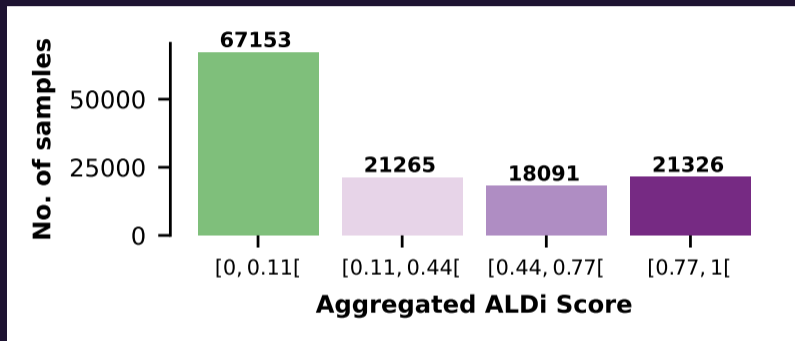
1 Labels into numeric values **2** Algebraic Mean



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

 Krippendorff's α (interval) = 0.63

ALDi scores Distribution



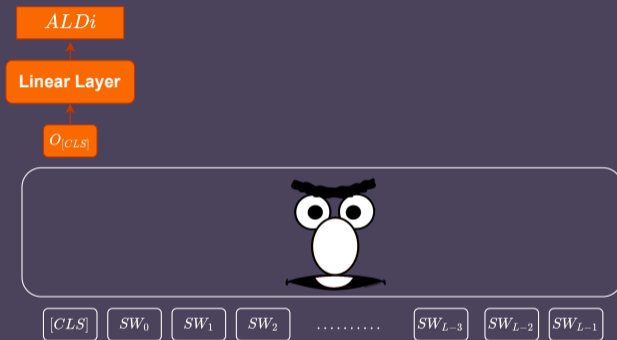
☆☆☆ Our Operationalization ☆☆☆

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

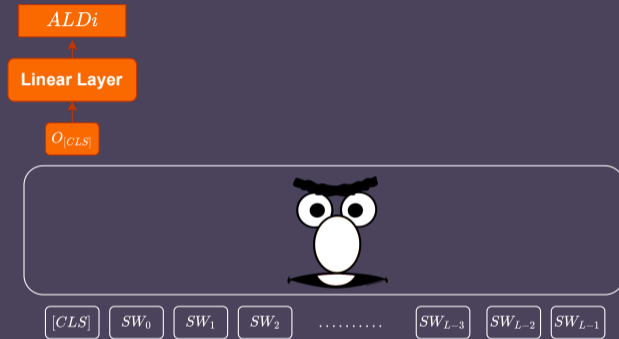


Building a model to estimate ALDi automatically

Sentence-ALDi model

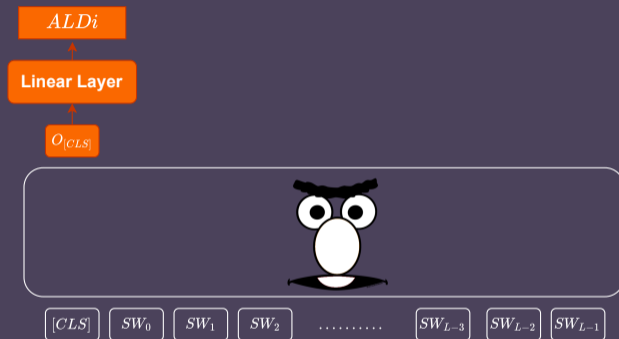


Sentence-ALDi model



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/>

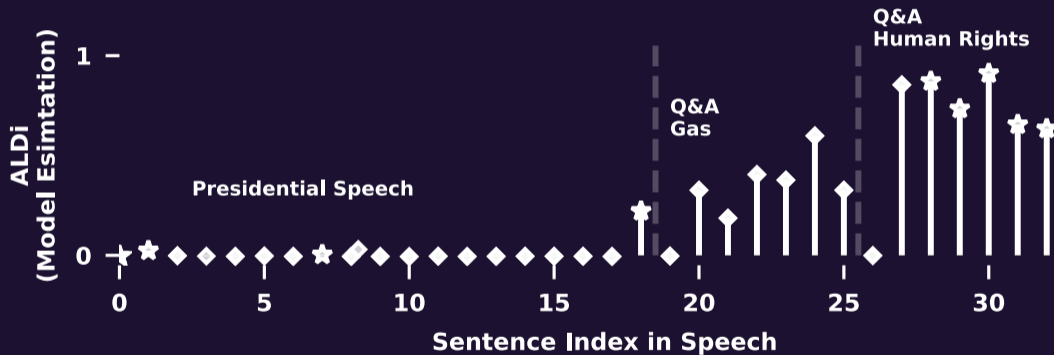
Egyptian President (El-Sisi) - 18/07/2022



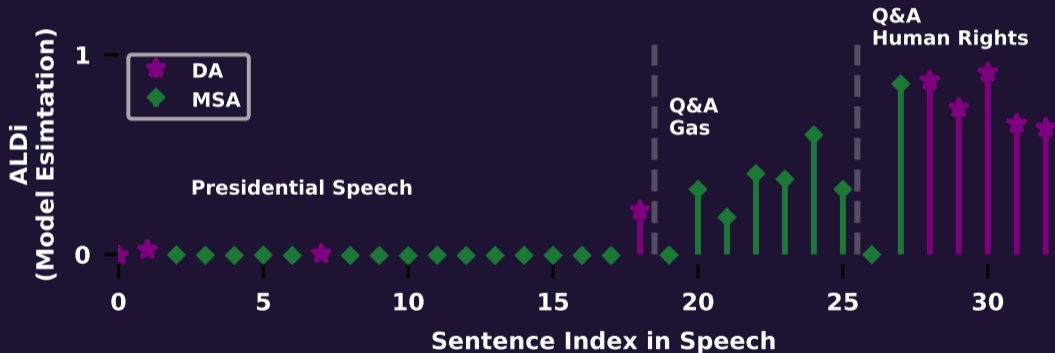
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Applications of ALDi

2 Impact of Interannotator Agreement

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

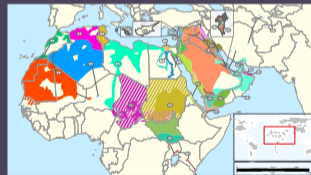
2) Annotating Multi-Dialect Arabic Datasets

Common Practice: 🎲 randomly assign to Arabic speakers

☢️☢️☢️ **Annotator's dialect \neq 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

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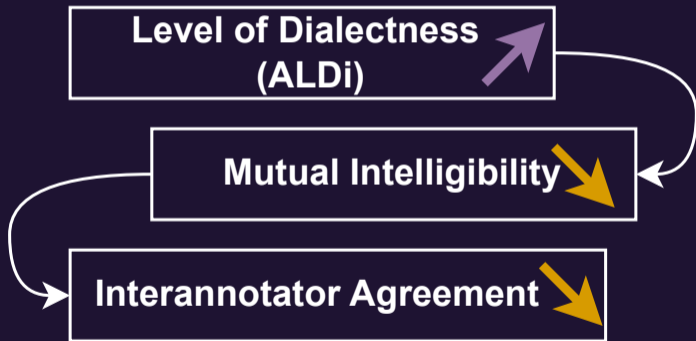
**Hard to crowdsource speakers of some dialects
(i.e., Limited resource  )**
(Mubarak and Darwish, 2016)



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:

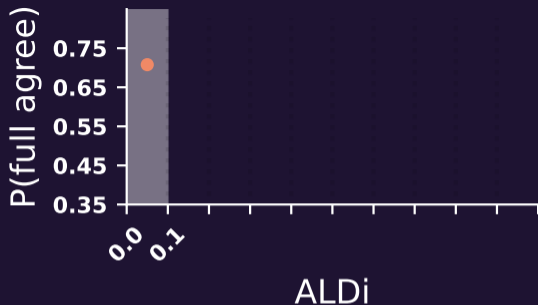
- 1 Estimate ALDi of samples.
- 2 Bin samples.
- 3 $P_{\text{bin}}(\text{Full Agreement})$

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



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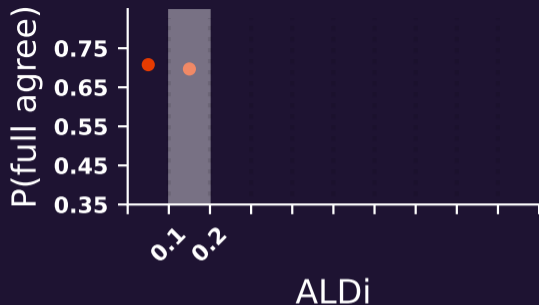


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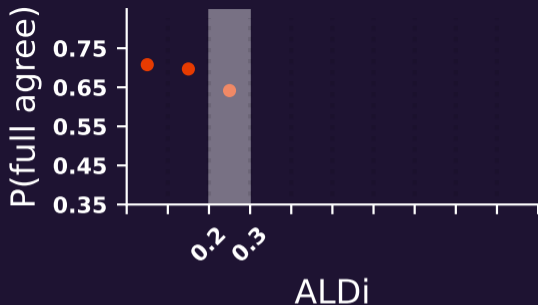


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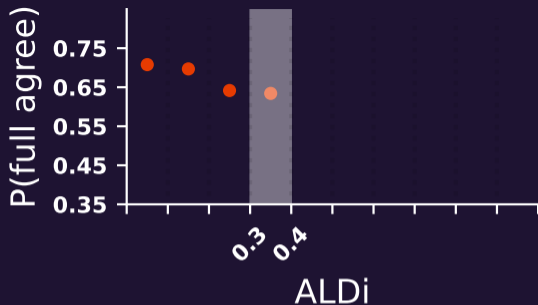


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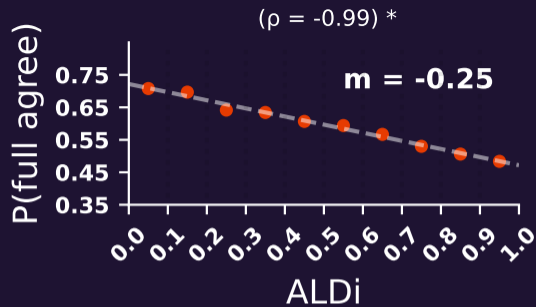


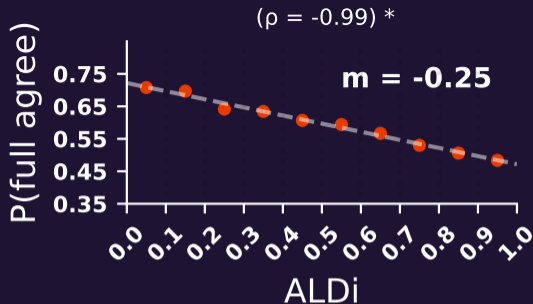
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Finding (1) - For 8 of 12 **non Dialect Identification** datasets

ALDi



Interannotator Agreement



with significant strong negative $\rho < -0.7$

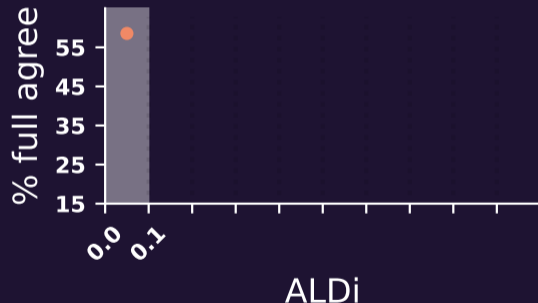
IAA - Dialect Identification Dataset

i Labels (Macro-regional):
MSA, Maghreb, Egypt, Levant, Gulf



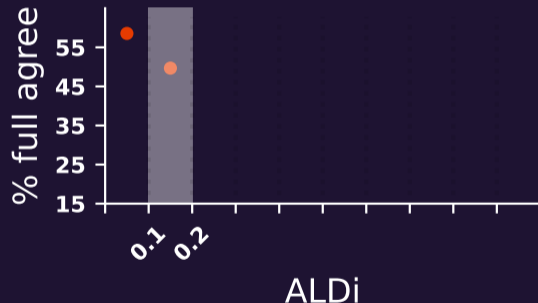
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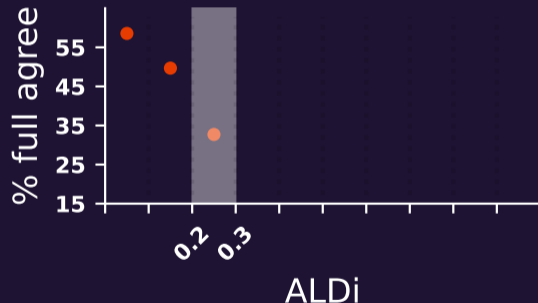
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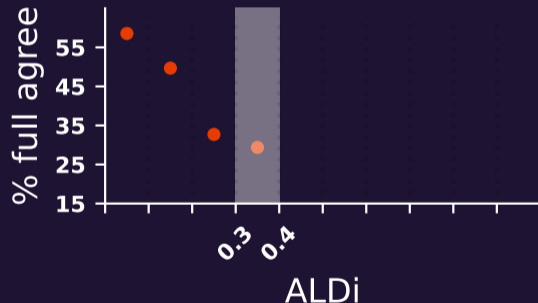
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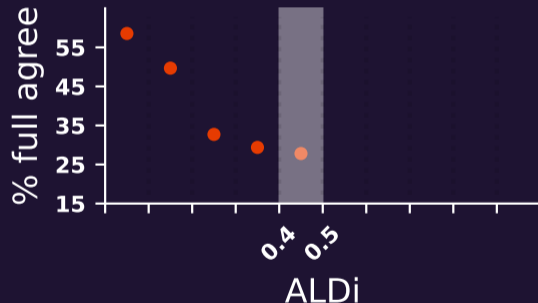
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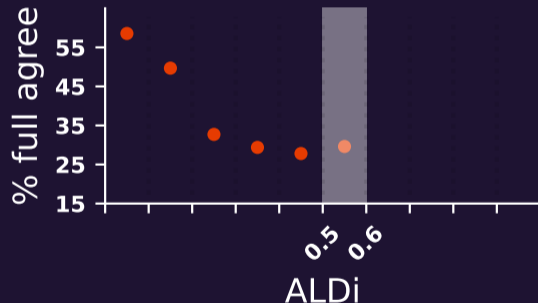
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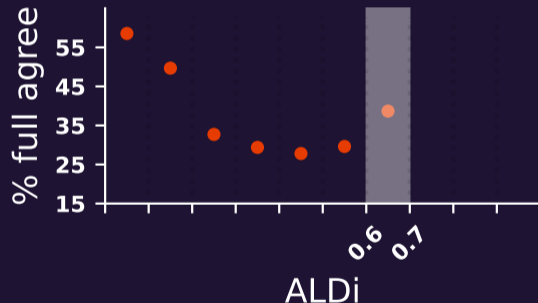
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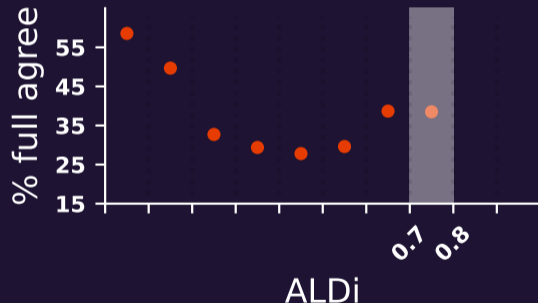
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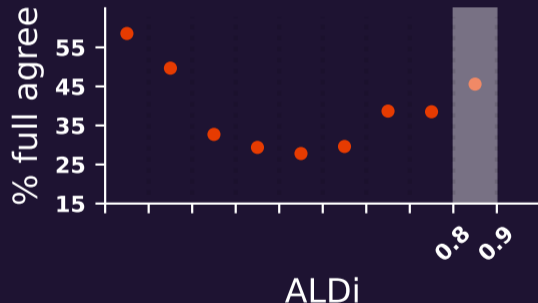
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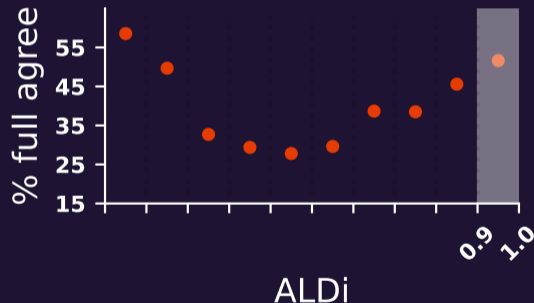
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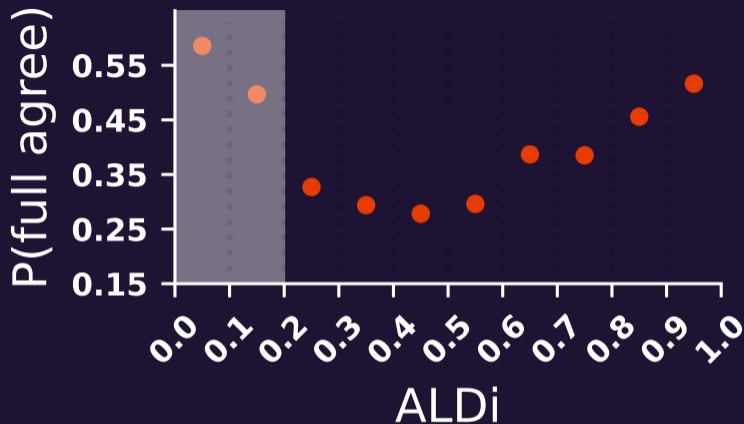
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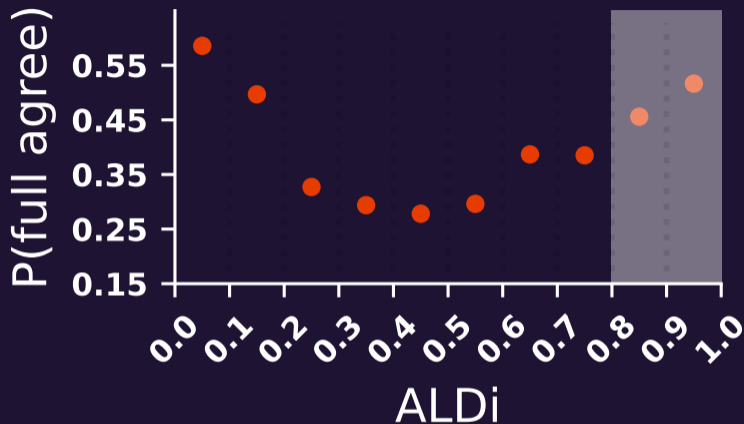
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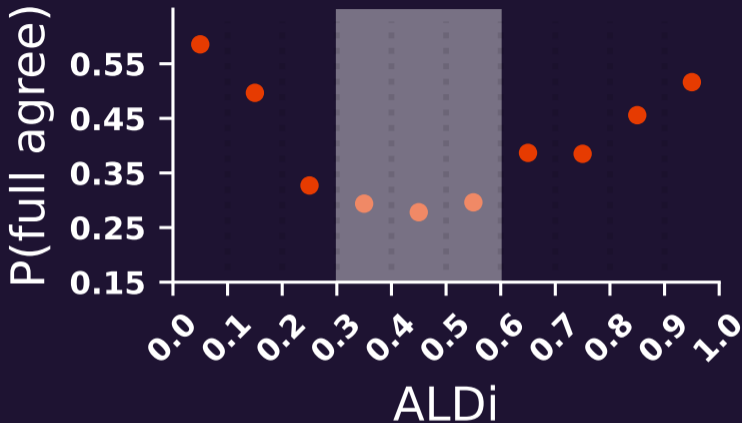






 MSA samples



 **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

-  Abdul-Mageed, Muhammad et al. (2024). "NADI 2024: The Fifth Nuanced Arabic Dialect Identification Shared Task." In: *Proceedings of The Second Arabic Natural Language Processing Conference (ArabicNLP 2024)* (ArabicNLP 2024).

References II

-  Abu Farha, Ibrahim and Walid Magdy (Dec. WANLP 2022).
“The Effect of Arabic Dialect Familiarity on Data Annotation.”
In: *Proceedings of the Seventh Arabic Natural Language Processing Workshop (WANLP)*. Ed. by Houda Bouamor et al.
Abu Dhabi, United Arab Emirates (Hybrid): Association for Computational Linguistics, pp. 399–408. DOI:
10.18653/v1/2022.wanlp-1.39. URL:
<https://aclanthology.org/2022.wanlp-1.39>.
-  Badawi, As-Said Muḥámmad (1973). *Levels of Contemporary Arabic in Egypt* (مستويات العربية المعاصرة في مصر). Dar Al-Maarif.

References III



Bergman, A. and Mona Diab (May ACL (findings) 2022).
“Towards Responsible Natural Language Annotation for the Varieties of Arabic.” In: *Findings of the Association for Computational Linguistics: ACL 2022*. Ed. by Smaranda Muresan, Preslav Nakov, and Aline Villavicencio. Dublin, Ireland: Association for Computational Linguistics, pp. 364–371. DOI: 10.18653/v1/2022.findings-acl.31. URL: <https://aclanthology.org/2022.findings-acl.31>.

References IV




Elfardy, Heba and Mona Diab (May 2012). "Simplified guidelines for the creation of Large Scale Dialectal Arabic Annotations." In: *Proceedings of the Eighth International Conference on Language Resources and Evaluation (LREC'12)* (LREC 2012). Istanbul, Turkey: European Language Resources Association (ELRA), pp. 371–378. URL: http://www.lrec-conf.org/proceedings/lrec2012/pdf/815_Paper.pdf.




Ferguson, Charles A (1959). "Diglossia." In: *word* 15.2, pp. 325–340.

References V

-  Habash, Nizar et al. (2008). "Guidelines for annotation of Arabic dialectness." In: *Proceedings of the LREC Workshop on HLT & NLP within the Arabic world* (Workshop on HLT & NLP - LREC 2008), pp. 49–53.

References VI


-  Keleg, Amr (May 2025). "LLM Alignment for the Arabs: A Homogenous Culture or Diverse Ones." In: *Proceedings of the 3rd Workshop on Cross-Cultural Considerations in NLP (C3NLP 2025)* (C3NLP 2025 - NAACL 2025). Ed. by Vinodkumar Prabhakaran et al. Albuquerque, New Mexico: Association for Computational Linguistics, pp. 1–9. ISBN: 979-8-89176-237-4. URL: <https://aclanthology.org/2025.c3nlp-1.1/>.

References VII



Keleg, Amr, Sharon Goldwater, and Walid Magdy (Dec. 2023).
“ALDi: Quantifying the Arabic Level of Dialectness of Text.” In:
*Proceedings of the 2023 Conference on Empirical Methods in
Natural Language Processing (EMNLP 2023)*. Ed. by
Houda Bouamor, Juan Pino, and Kalika Bali. Singapore:
Association for Computational Linguistics, pp. 10597–10611.
DOI: 10.18653/v1/2023.emnlp-main.655. URL:
<https://aclanthology.org/2023.emnlp-main.655>.

References VIII

-  Keleg, Amr and Walid Magdy (Dec. 2023). “Arabic Dialect Identification under Scrutiny: Limitations of Single-label Classification.” In: *Proceedings of ArabicNLP 2023* (ArabicNLP 2023). Ed. by Hassan Sawaf et al. Singapore (Hybrid): Association for Computational Linguistics, pp. 385–398. DOI: 10.18653/v1/2023.arabichlp-1.31. URL: <https://aclanthology.org/2023.arabichlp-1.31>.

References IX





Keleg, Amr, Walid Magdy, and Sharon Goldwater (Aug. 2024).
“Estimating the Level of Dialectness Predicts Inter-annotator Agreement in Multi-dialect Arabic Datasets.” In: *Proceedings of the 62nd Annual Meeting of the Association for Computational Linguistics (Volume 2: Short Papers)* (ACL 2024). Ed. by Lun-Wei Ku, Andre Martins, and Vivek Srikumar. Bangkok, Thailand: Association for Computational Linguistics, pp. 766–777. DOI: 10.18653/v1/2024.acl-short.69. URL: <https://aclanthology.org/2024.acl-short.69/>.

References X

-  Lahlali, M (2011). "The Arab Spring and the discourse of desperation: shifting from an authoritarian discourse to a "democratic one"." In: *The Journ. of Arab Media and Society.—Cairo: American Univ. in Cairo* 13.
-  Mubarak, Hamdy and Kareem Darwish (2016). "Demographic surveys of Arab annotators on CrowdFlower." In: *Proceedings of ACM WebSci16 Workshop "Weaving Relations of Trust in Crowd Work: Transparency and Reputation across Platforms*.

References XI

-  S'hiri, Sonia (2002). "Speak Arabic please!: Tunisian Arabic speakers' linguistic accommodation to middle Easterners." In: *Language contact and language conflict in Arabic: Variations on a sociolinguistic theme*, pp. 149–174.
-  Zaidan, Omar F. and Chris Callison-Burch (June 2011). "The Arabic Online Commentary Dataset: an Annotated Dataset of Informal Arabic with High Dialectal Content." In: *Proceedings of the 49th Annual Meeting of the Association for Computational Linguistics: Human Language Technologies*. Portland, Oregon, USA: Association for Computational Linguistics, pp. 37–41. URL: <https://aclanthology.org/P11-2007>.

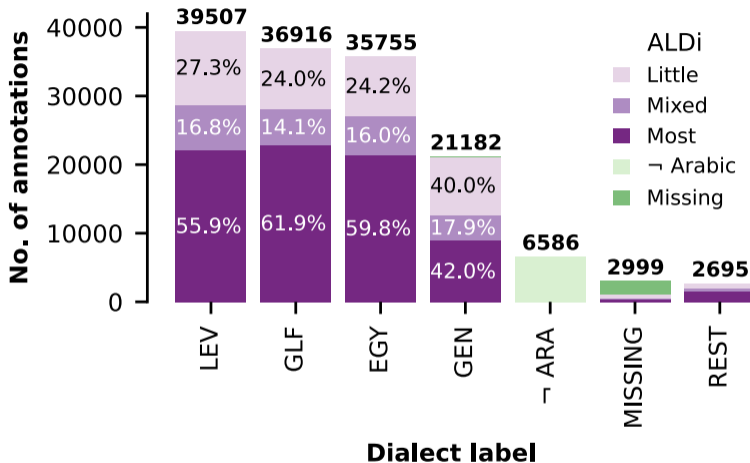
Exposure is key for Intelligibility

“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)

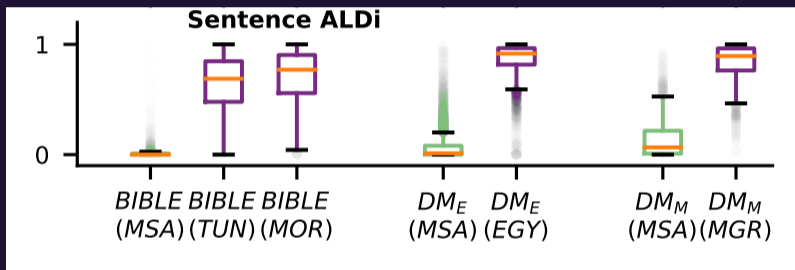
Sometimes compared to Romance Languages.

Why not different languages?

“MSA is a kind of **communally-owned reservoir** that Arabs use to ORANGEmake 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?

- 1,120 sentences.
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- 3 annotators from 9 different countries (total of 27)

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

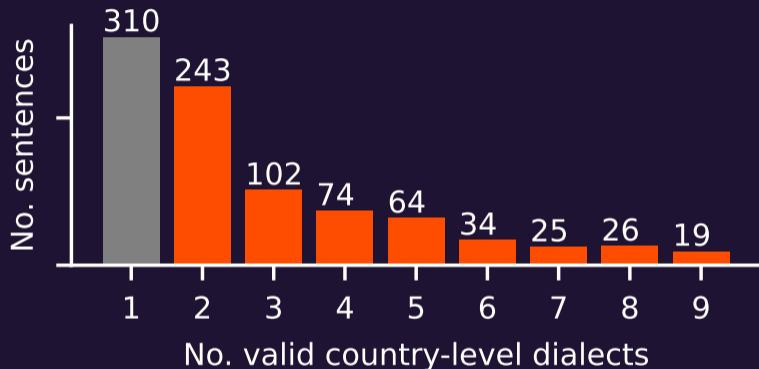
L3 Informal (or Vulgar) Colloquial

Note: The levels and their descriptions were provided in Arabic.

Country	N valid	Krip. α
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

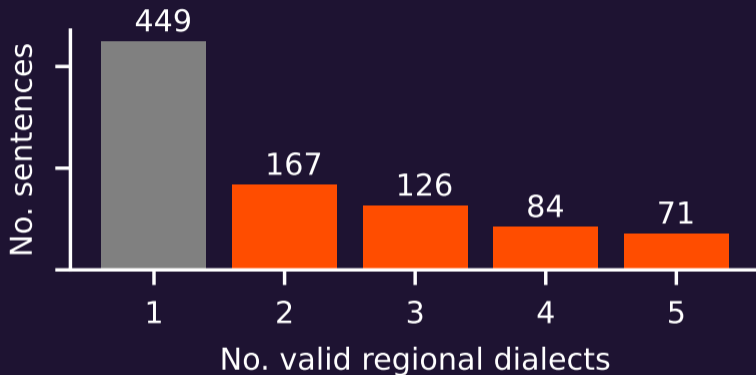
✓ Improved alpha scores than AOC-ALDi.

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!