

# Guidelines for Fine-grained Sentence-level Arabic Readability Annotation

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## Abstract

This paper presents the annotation guidelines of the Balanced Arabic Readability Evaluation Corpus (**BAREC**), a large-scale resource for fine-grained sentence-level readability assessment in Arabic. **BAREC** includes 69,441 sentences (1M+ words) labeled across 19 levels, from kindergarten to postgraduate. Based on the Taha/Arabi21 framework, the guidelines were refined through iterative training with native Arabic-speaking educators. We highlight key linguistic, pedagogical, and cognitive factors in determining readability and report high inter-annotator agreement: Quadratic Weighted Kappa 81.8% (substantial/excellent agreement) in the last annotation phase. We also benchmark automatic readability models across multiple classification granularities (19-, 7-, 5-, and 3-level). The corpus and guidelines are publicly available.<sup>1</sup>

## 1 Introduction

Text readability plays a crucial role in comprehension, retention, reading speed, and engagement (DuBay, 2004). When texts exceed a reader’s ability, they can lead to frustration and disengagement (Klare, 1963). Readability is shaped by both the content and presentation (Nassiri et al., 2023). In educational settings, readability leveling is widely used to align texts with students’ reading abilities, promoting independent and more effective learning (Allington et al., 2015; Barber and Klauda, 2020).

Fine-grained readability systems, like Fountas and Pinnell’s 27-level scale in English (Fountas and Pinnell, 2006), and Taha’s 19-level Arabic system (Taha-Thomure, 2017), guide progression from early readers to adult fluency. These levels support instructional goals and can be mapped to broader categories for practical use in NLP.

We present the Balanced Arabic Readability Evaluation Corpus (**BAREC**), a large-scale dataset

RL Grade	Example
1 KG	Ball كُرَة
3 1st	The bedroom غُرْفَةُ النَّوْمِ
6 2nd	My behavior is my responsibility سُلُوكِي مَسْئُولِيَّتِي
10 4th	كانت الحديقة واسعة، تطل على شاطئ النيل، The garden was spacious, overlooking the Nile.
14 8th	تعريف أصول الفقه Definition of Islamic Jurisprudence Principles
17 Uni	بين طعن القنا وحقق البُؤود Between lance thrusts and ensign flutters

Table 1: Examples by Reading Level (RL) and grade.

of 69K+ sentences<sup>2</sup> (1M+ words) across a broad space of genres and 19 readability levels. Based on the Taha/Arabi21 framework (Taha-Thomure, 2017), which has been instrumental in tagging over 9,000 children’s books, **BAREC** guidelines enable standardized, sentence-level readability evaluation across diverse genres and educational levels, ranging from kindergarten to postgraduate comprehension (see Table 1). Our contributions are as follows:

- We **define detailed annotation guidelines** for Arabic sentence-level readability across a fine-grained 19-level scale.
- We **apply and refine these guidelines** through annotation of a diverse, large-scale corpus, analyzing annotator agreement and sources of difficulty in this nuanced task.
- We **build and evaluate readability models** across multiple granularities (19, 7, 5, and 3 levels) to provide baseline results for various research and application needs.

Next, §2 reviews related work, §3 outlines the annotation framework, §4 covers data selection, and §5 discusses evaluation results.

<sup>1</sup><http://barec.camel-lab.com>

<sup>2</sup>We use *sentence* to refer to syntactic sentences as well as shorter standalone text segments (e.g., phrases or titles).

Authors	Project	Metric	Levels	Unit	Size	Content
Al-Khalifa and Al-Ajlan (2010)	Arability	Readability	3	Document	150	School Textbooks
Forsyth (2014)	DLI Corpus	ILR	5 (3)	Document	179	L2 Learner
Kilgarriff et al. (2014)	KELLY	CEFR	6	Word	9,000	Most Frequent
Taha-Thomure (2017)	Taha/Arabi21	Readability	19	Document	9,000	Children’s Books
Al Khalil et al. (2020)	SAMER Lexicon	Readability	5	Word	40,000	General Vocab
Habash and Palfreyman (2022)	ZAEBUC	CEFR	6	Document	214	Prompted Essays
Naous et al. (2024)	ReadMe++	CEFR	6	Sentence	1,945	Multi-domain
Soliman and Familiar (2024)	Arabic Vocab Profile	CEFR	2	Word	1,200	L2 Learner (A1, A2)
El-Haj et al. (2024)	DARES	Grade Level	12	Sentence	13,335	School Textbooks
Alhafni et al. (2024)	SAMER Corpus	Readability	3	Word	159,265	Literature
Bashendy et al. (2024)	QAES	AES	7×5	Document	195	Argumentative Essays
Our Work	BAREC	Readability	19 (7–5–3)	Sentence	69,441	Multi-domain

Table 2: Overview of Arabic readability and proficiency-related corpora.

## 2 Related Work

**Automatic Readability Assessment** Automatic readability assessment has been widely studied, resulting in numerous datasets and resources (Collins-Thompson and Callan, 2004; Pitler and Nenkova, 2008; Feng et al., 2010; Vajjala and Meurers, 2012; Xu et al., 2015; Xia et al., 2016; Nadeem and Ostendorf, 2018; Vajjala and Lučić, 2018; Deutsch et al., 2020; Lee et al., 2021). Early English datasets were often derived from textbooks, as their graded content naturally aligns with readability assessment (Vajjala, 2022). However, copyright restrictions and limited digitization have driven researchers to crowdsource readability annotations from online sources (Vajjala and Meurers, 2012; Vajjala and Lučić, 2018) or leverage CEFR-based L2 assessment exams (Xia et al., 2016).

**Arabic Readability Efforts** Arabic readability research has explored text leveling and assessment in multiple frameworks (Nassiri et al., 2023).

Taha-Thomure (2017) proposed a 19-level Arabic text leveling framework for educators, inspired by Fountas and Pinnell (2006) and focused on children’s literature. Targeting full texts (books), particularly for early education, with 11 of the 19 levels covering up to 4th grade, the system supports teachers in matching books to students’ reading abilities. Taha-Thomure (2017)’s procedural framework outlines ten qualitative and quantitative criteria: text genre, abstractness of ideas, vocabulary and its proximity to dialects, text authenticity, book production quality, content suitability, sentence structure, illustrations, use of diacritics, and word count. The Arab Thought Foundation adopted this framework under its Arabi21 initiative, which funded the leveling of over 9,000 children’s books.

Other efforts applied CEFR leveling to Arabic, including the KELLY project’s frequency-based word lists, manually annotated corpora such as ZAEBUC (Habash and Palfreyman, 2022) and ReadMe++ (Naous et al., 2024), and vocabulary profiling (Soliman and Familiar, 2024). El-Haj et al. (2024) introduced DARES, a readability assessment dataset collected from Saudi school materials. The SAMER project (Al Khalil et al., 2020) developed a lexicon with a five-level readability scale, leading to the first manually annotated Arabic parallel corpus for text simplification (Alhafni et al., 2024). Bashendy et al. (2024) presented a corpus of Arabic essays annotated across organization and style traits.

Automated readability assessment in Arabic has evolved from rule-based models using surface features (Al-Dawsari, 2004; Al-Khalifa and Al-Ajlan, 2010) to machine learning approaches with POS, morphology (Forsyth, 2014; Saddiki et al., 2018), and script features like OSMAN (El-Haj and Rayson, 2016). Recent work (Liberato et al., 2024) shows strong results with pretrained models on the SAMER corpus.

**Our Approach** Building on prior work, we curated the BAREC corpus across diverse genres and readability levels, manually annotating it at the sentence level using adapted Taha/Arabi21 guidelines (Taha-Thomure, 2017). Sentence-level annotation balances the coarse granularity of document-level labels and the limited context of word-level labels. This allows finer control and more objective assessment of textual variation. Table 2 compares BAREC with earlier efforts. To our knowledge, BAREC is the largest and most fine-grained manually annotated Arabic readability resource.



RL	Arabic Sentence/Phrase	Translation	Reasoning
1-alif	أَرْنَبِ <b>Rabbit</b>		One bisyllabic familiar noun
2-ba	مَلْعَبٌ وَاسِعٌ <b>A large playground</b>		Noun-adjective
3-jim	أنا أحب اللون الأحمر. I love <b>the</b> color red.		Definite article
4-dal	الشمس تشرق في الصباح الباكر. The sun rises early <b>in the morning</b> .		Prepositional phrase
5-ha	القطعة تستريح على السرير وتستمتع بأشعة الشمس الدافئة. The cat rests on the bed <b>and enjoys the warm sunshine</b> .		A conjoined sentence
6-waw	سلوكي <b>مُسَوِّلِيَّتِي</b> My behavior is <b>my responsibility</b>		Five syllable word
7-zay	الأصدقاء يحتفلون بعيد ميلاد صديقهم بكعكة وهدايا رائعة. <b>Friends</b> celebrate their friend's birthday with cake and amazing gifts.		Broken plural
8-ha	أستمعُ إلى كُلِّ فقرةٍ مِنَ الفقرَتَيْنِ الاتينيتين، ثُمَّ أَجِيبُ: <b>then</b> I answer:	I listen to each of the following two paragraphs, <b>then</b> I answer:	ح (then) is in level 8-ha
9-ta	وقال بكلام فصيح مزعج: <b>يا سمك يا سمك</b> هل أنت على العهد القديم مقيم <b>fish</b> , do you abide by the old promise	He said in annoying, eloquent words: <b>Oh fish, oh fish</b> , do you abide by the old promise	Vocative construction
10-ya	وسألتك هل كنتُ تَتَّهمُونَهُ بالكذب قبل أن يقول ما قال فذكرتُ أن لا، <b>you were</b> accusing him of lying before he said what he said, and you said no.		Auxiliary Kaana
11-kaf	حسام سعيدٌ قلبه بسبب فوز فريقه. Hossam, his <b>heart is happy</b> because of his team's victory.		Acting derivative (happy is predicative)
12-lam	لا أحد يجمع هذه الزهور معًا في باقة، فهي منتشرة جدًا — <b>حتى إنه كان من المعروف to grow between paving stones, and spring up everywhere like weeds</b> — and they have the very unsightly name of “dog-flowers” or “dandelions.”	No one puts these flowers together in a bouquet, they are so common — <b>they have even been known to grow between paving stones, and spring up everywhere like weeds</b> — and they have the very unsightly name of “dog-flowers” or “dandelions.”	Parenthetical phrase
13-mim	ومن يفعل المعروف مع غير أهله يجازي كما <b>جوزي مجير أم عامر underserving will be rewarded like he who gave shelter to a hyena</b>	<b>And whoever offers good deeds to someone underserving will be rewarded like he who gave shelter to a hyena</b>	Conditional phrase
14-nun	حيث إن هذه الزيادة في <b>الجسيمات المشحونة</b> تشير إلى خروج المركبة من نطاق تأثير الرياح الشمسية الذي يسمى الغلاف الشمسي (والذي يعتبر حسب بعض التعاريف حدود المجموعة الشمسية). This increase in <b>charged particles</b> indicates the spacecraft's departure from the influence of the <b>solar wind</b> , which is called <b>the heliosphere</b> (which, according to some definitions, is the border of the <b>solar system</b> ).		General geography vocabulary
15-sin	وكان من عادتها أن تقارن بينها وبين بطلة الرواية إذا أحسّت منه إعجابًا بها أو ثناء عليها، وتسأله في ذلك أسئلة ذكية خبيثة لا تسهل <b>المغالطة في جوابها</b> ، إلا على سبيل المزاح <b>did not allow answering deceptively</b> , except by joking and teasing.	It was her habit to compare herself with the heroine of the novel when she felt his admiration or praise for her, asking him smart and tricky questions <b>that did not allow answering deceptively</b> , except by joking and teasing.	Specialized vocabulary that requires understanding the concept to comprehend its use
16-ayn	ويذهب المؤرخون إلى أن النابغة الذبياني كان من <b>المحكمين</b> ، تقام له في هذه الأسواق قبة يذهب إليها الشعراء ليعرضوا شعرهم، فمن أشاد به ذاع صيته، وتناقلت شعره الركبان. Historians assert that <b>Al-Nabigha Al-Dhubyani</b> was one of the <b>arbiters</b> . In these markets, a dome is erected for him where poets go to present their poetry. Whomever he praised, <b>his fame spread</b> , and his poetry circulated among the <b>caravans</b> .		Specialized and uncommon vocabulary
17-fa	بين طلعن القتا وخفق البتود Between the thrusts of <b>lances</b> and the fluttering of <b>ensigns</b>		Heritage vocabulary familiar to a novice specialist
18-sad	إلا الأوارئ لآيا ما أبيتها والنوى كالحوض بالمظلومة الجد <b>I wasn't able to see except with extreme effort and difficulty like a water basin in solid undrillable land</b>		Specialist vocabulary; symbolic poetic ideas requiring prior knowledge
19-qaf	كان حنوج المالكية غوةً خلايا سفين <b>leaving the Dadi valley were great ships</b> بالتواصف من دد	As if <b>the camel saddles of the Malikiyya caravan leaving the Dadi valley were great ships</b>	Advanced specialist vocabulary; symbolic poetic ideas requiring prior knowledge

Table 3: Representative subset of examples of the 19 BAREC readability levels, with English translations, and readability level reasoning. Underlining is used to highlight the main keys that determined the level.

### 3.4 Dimensions of Textual Features

To determine the BAREC level, we define six textual dimensions that identify key features necessary to unlock each level:

**1. Number of Words** Counts unique printed words (ignoring punctuation and diacritics). Used only up to level **11-kaf** (max 20 words).

**2. Orthography & Phonology** Focuses on word length (syllables) and letters like Hamzas. Final

diacritics are ignored (words read in *waqf*), e.g., **أَرْنَبِ** *Ar.nabū* ‘rabbit’ has 2 syllables: *ar-nab*.

**3. Morphology** Covers derivation and inflection (tense, voice, number, etc.). Simpler forms appear at lower levels (e.g., present tense before past, singular before plural). Used up to level **13-mim**.

**4. Syntactic Structures** Tracks sentence complexity, from single words (**1-alif**) to complex constructions. Used up to level **15-sin**.

**5. Vocabulary** Central at all levels. Overlapping dialect and MSA vocabulary appear at easier levels; technical terms are introduced at harder levels. Arabized foreign words are treated as part of the language, while non-Arabic script is excluded.

**6. Ideas & Content** Evaluates needed prior knowledge, symbolic unpacking, and conceptual linking. Levels progress from familiar to specialized knowledge and from literal to abstract ideas. We recognize that such evaluations are complex and may vary subjectively among readers within the same age or education group.

**Problems and Difficulties** Annotators are instructed to report issues such as spelling errors, colloquial language, or sensitive topics. Difficulty is noted when annotations cannot be made due to conflicting guidelines.

The **BAREC** pyramid (Figure 1) illustrates which aspects are used (broadly) for which levels. For example, spelling criteria are only used up to level **7-zay**, while syntax is used until level **15-sin**, and word count is not used beyond level **11-kaf**. A full set of examples with explanations of leveling choices is in Table 3. The *Annotation Cheat Sheet* used by the annotators in Arabic and its translation in English are included in Appendix A. The full guidelines are publicly available.<sup>1</sup> For more on Arabic linguistic features, see [Habash \(2010\)](#).

### 3.5 Annotation Process

**Sentence Segmentation** Since our starting point is a text excerpt, typically a paragraph or two (~500±200 words) from each source, we begin with sentence-level segmentation and initial text flagging. We followed the Arabic sentence segmentation guidelines by [Habash et al. \(2022\)](#).

**Sentence Readability Annotation** Each annotator is presented with a batch of 100 randomly selected sentences to annotate. The annotation was done through a simple Google Sheet interface (see Appendix A.3), which provides details such as sentence word count, and the guidelines constraints for the selected level to provide feedback confirmation to the annotator. The annotators are instructed to follow this procedure: **First** they read the sentence and make sure it has no flaws that can lead to excluding it. **Second**, they think about the meaning of the sentence noting any ambiguities due to diacritic absence or limited context, and consciously decide on the simpler reading in case of

multiple readings. **Third**, they make an initial assessment of the lowest possible level based on word count. **Fourth**, they look for specific phenomena that allow increasing the level to the highest possible. For example, the sixth sentence in Table 3, *سلوكي مسؤوليتي slwky ms'wlyty* ‘my behavior is my responsibility’ has two words, which automatically sets it as level **2-ba** or higher. The presence of the first person pronominal clitic *ي+* elevates the level to **3-jim**; however, the fact that the second word has five syllables raises the level further to **6-waw**. No other keys can take it higher.

Annotation averaged 2.5 hours per 100-sentence batch (1.5 minutes per sentence), reflecting the careful and rigorous approach taken by annotators to ensure high-quality, consistent labeling across a diverse and challenging dataset.

### 3.6 Annotation Team

The **BAREC** annotation team included six native Arabic-speaking educators (A0-A5), most with advanced degrees in Arabic Literature or Linguistics. A0 had prior experience in computational linguistics annotation, while A1-A5 brought extensive expertise in readability assessment from the Taha/Arabi21 project. A0 handled sentence segmentation and initial text selection; and A5 led the annotation team in assigning readability labels. Annotator profiles, covering demographic, educational, linguistic, and teaching backgrounds, are listed in Appendix A.4.

### 3.7 Training and Quality Control

Annotators A1-A5 received thorough training, including three shared pilot rounds that enabled in-depth discussion and refinement of the guidelines.

To ensure consistency, the initial 10,658 sentences (Phase 1) were double-reviewed before annotating the full 69K (1M+ words). Inter-annotator agreement (IAA) was assessed on 19 blind batches (excluding pilots 1 and 2), followed by group unification to support quality control and prevent drift. Only unified labels appear in the official release. The multiple IAA annotations will be released separately to support research on readability annotations.<sup>1</sup> Details on IAA are in Section 5.3).

In total, the annotators labeled 92.6K sentences; 25% were excluded from the final corpus: 3.3% were problematic (typos and offensive topics), 11.5% from early double annotations, and 10.3% from IAA rounds (excluding unification).



Category	Domain	Foundational	Advanced	Specialized	All
Documents	Arts & Humanities	562 (29%)	478 (25%)	327 (17%)	<b>1,367 (71%)</b>
	Social Sciences	44 (2%)	168 (9%)	163 (8%)	<b>375 (20%)</b>
	STEM	27 (1%)	85 (4%)	68 (4%)	<b>180 (9%)</b>
	All	<b>633 (33%)</b>	<b>731 (38%)</b>	<b>558 (29%)</b>	<b>1,922 (100%)</b>
Sentences	Arts & Humanities	24,978 (36%)	15,285 (22%)	10,179 (15%)	<b>50,442 (73%)</b>
	Social Sciences	2,270 (3%)	5,463 (8%)	6,586 (9%)	<b>14,319 (21%)</b>
	STEM	533 (1%)	1,948 (3%)	2,199 (3%)	<b>4,680 (7%)</b>
	All	<b>27,781 (40%)</b>	<b>22,696 (33%)</b>	<b>18,964 (27%)</b>	<b>69,441 (100%)</b>
Words	Arts & Humanities	274,497 (26%)	222,933 (21%)	155,565 (15%)	<b>652,995 (63%)</b>
	Social Sciences	26,692 (3%)	110,226 (11%)	138,813 (13%)	<b>275,731 (27%)</b>
	STEM	12,879 (1%)	48,501 (5%)	49,265 (5%)	<b>110,645 (11%)</b>
	All	<b>314,068 (30%)</b>	<b>381,660 (37%)</b>	<b>343,643 (33%)</b>	<b>1,039,371 (100%)</b>

Table 4: **BAREC** corpus statistics in documents, sentences, and words, across domain and readership levels.

## 4 BAREC Corpus

### 4.1 Corpus Selection

In the process of corpus selection, we aimed to cover a wide educational span as well as different domains and topics. We collected the corpus from 1,922 documents, which we manually categorized into three domains: **Arts & Humanities**, **Social Sciences**, and **STEM**,<sup>4</sup> and three readership groups: **Foundational**, **Advanced**, and **Specialized**.<sup>5</sup> Table 4 shows the distribution of the documents, sentences and words across domains and groups. The corpus emphasizes educational coverage, with a higher-than-usual proportion of foundational-level texts. Domain variation reflects text availability and reader interest (more Arts & Humanities, less STEM). Texts were sourced from 30 resources, all either public domain, within fair use, or used with permission. Some were selected due to existing annotations. Notably, 25% of sentences came from new sources that were manually digitized. See Appendix C for resource details.

### 4.2 Readability Statistics

Figure 2 shows sentence distribution across **BAREC**-19 levels and their mappings to coarser levels (7, 5, and 3). The distribution is uneven, with 63% of sentences in the middle levels (**10-ya**~fourth grade to **14-nun**~ninth grade) reflecting natural text complexity and real-world usage.

<sup>4</sup>**Arts & Humanities**: literature, philosophy, religion, education, and related news. **Social Sciences**: business, law, social studies, education, and related news. **STEM**: science, technology, engineering, math, education, and related news.

<sup>5</sup>**Foundational**: Learners up to 4th grade (age 10), focused on basic literacy skills. **Advanced**: Adult readers with average abilities, handling moderate complexity texts. **Specialized**: Advanced readers (typically 9th grade+), engaging with domain-specific texts.

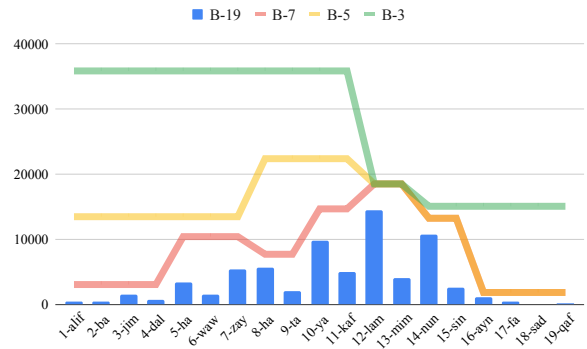


Figure 2: The distribution of sentences across **BAREC**-19 levels (blue), and their mapping to coarser levels.

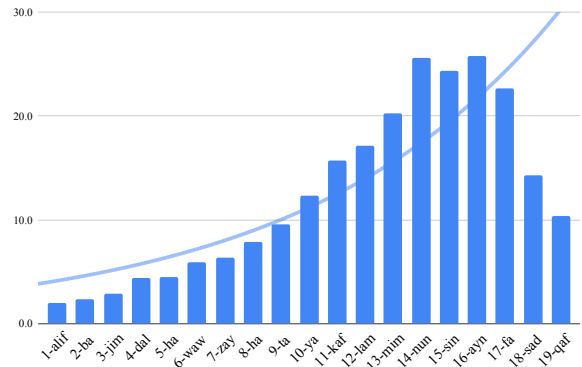


Figure 3: The average sentence word count across **BAREC**-19 levels, with trend line.

Figure 3 shows average sentence length by level, which correlates strongly with readability (Pearson  $r=81\%$ ). The drop at higher levels may result from shorter classical poetry lines.

Figure 4 shows *relative* distribution of readership groups and domains across readability levels. Foundational texts dominate lower levels and specialized texts higher ones. STEM and Social Science texts have a higher relative appearance in the upper mid levels.

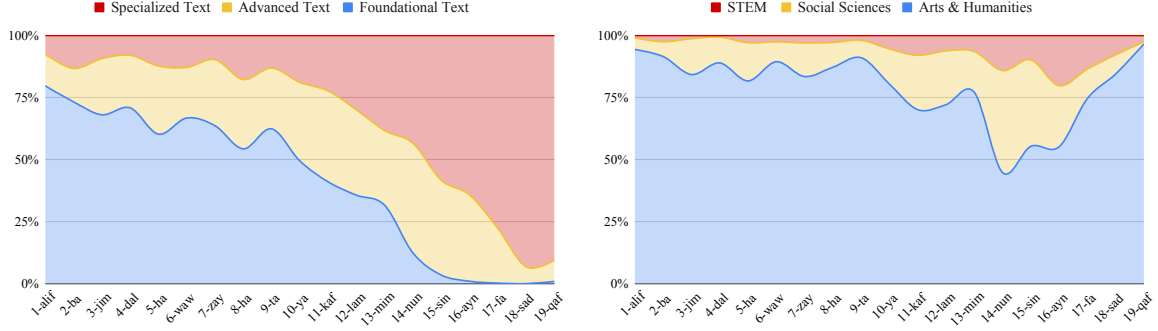


Figure 4: The relative distribution of readership groups and domains across **BAREC** levels.

## 5 Evaluation and Analysis

### 5.1 Metrics

We evaluate readability models and IAA using Accuracy, Adjacent Accuracy, Average Distance, and Quadratic Weighted Kappa (QWK), with QWK as our primary metric.

**Accuracy (Acc)** The percentage of cases where the predicted class matches the reference class in the 19-level scheme ( $\text{Acc}^{19}$ ), as well as three variants,  $\text{Acc}^7$ ,  $\text{Acc}^5$ , and  $\text{Acc}^3$ , which collapse the 19-level scheme into 7, 5, and 3 levels, respectively (Section 3.2).

**Adjacent Accuracy ( $\pm 1 \text{ Acc}^{19}$ )** The proportion of predictions that are either exactly correct or off by at most one level.

**Average Distance (Dist)** The average absolute difference between two sets of labels. For example, the distance between **2-ba** and **4-dal** is 2.

**Quadratic Weighted Kappa (QWK)** An extension of Cohen’s Kappa (Cohen, 1968; Doewes et al., 2023), measuring agreement between predicted and true labels, with a quadratic penalty for larger misclassifications.

### 5.2 Corpus Splits

We split the corpus at the document level into **Train** ( $\sim 80\%$ ), **Dev** ( $\sim 10\%$ ), and **Test** ( $\sim 10\%$ ). Sentences from IAA studies are distributed across splits. For resources with existing splits, such as CamelTB (Habash et al., 2022) and ReadMe++ (Naous et al., 2024), we adopted their original splits. Table 5 reports the splits by documents, sentences, and words. Due to IAA and external corpus constraints, final proportions slightly deviate from exact 80-10-10. See Appendix B for full and split readability level distributions.

Split	#Documents	#Sentences	#Words
Train	1,518 (79%)	54,845 (79%)	832,743 (80%)
Dev	194 (10%)	7,310 (11%)	101,364 (10%)
Test	210 (11%)	7,286 (10%)	105,264 (10%)
<b>All</b>	<b>1,922 (100%)</b>	<b>69,441 (100%)</b>	<b>1,039,371 (100%)</b>

Table 5: **BAREC** corpus splits.

Stage	#Sets	Distance	Acc <sup>19</sup>	$\pm 1 \text{ Acc}^{19}$	QWK
Pilot 3	1	1.69	37.5%	58.5%	79.3%
Phase 1	2	1.38	48.4%	64.4%	80.2%
Phase 2A	6	1.21	49.4%	67.4%	72.4%
Phase 2B	10	0.80	67.6%	78.3%	78.8%
Overall / Macro	19	1.04	58.2%	72.3%	76.9%
Phase 2 / Macro	16	0.96	60.8%	74.2%	76.4%
Phase 2 / Micro	16	0.95	61.1%	74.4%	81.8%

Table 6: Average pairwise inter-annotator agreement (IAA) across different annotation stages. Macro/Micro indicate the form of averaging, over sets or sentences, respectively. Phase 2 = Phase 2A and 2B.

### 5.3 Inter-Annotator Agreement (IAA)

**Pairwise Agreement** Table 6 summarizes results for 19 IAA sets (excluding Pilots 1 and 2). We observe steady improvement from Pilot 3 to Phase 2B, with reduced distance and higher accuracy. The overall macro-average QWK is 76.9%, indicating substantial agreement and suggesting that most disagreements are minor (Cohen, 1968; Doewes et al., 2023). In Phase 2, the final and largest phase, the micro-average QWK rises to 81.8%.

Figure 5 presents a confusion matrix of sentence-level pairwise agreements for Phase 2 IAA sentences, using F-scores to account for the unbalanced level distribution. The strong diagonal (exact matches) reflects a high degree of agreement, consistent with the overall IAA results. However, accuracy varies across levels, with more disagree-

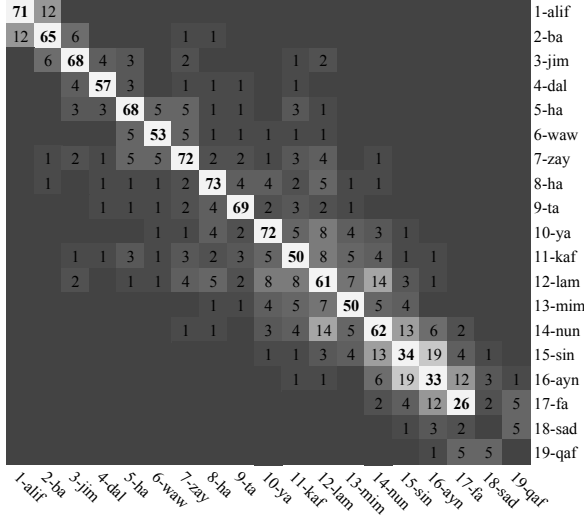


Figure 5: Confusion matrix for annotator pairwise agreement on Phase 2 IAA sentences normalized as F-scores.

ment at the harder higher levels. This may stem from the guidelines emphasizing vocabulary and content at the higher levels, features that are inherently more subjective than the textual feature cues used at lower levels.

**Unification Agreement** After each IAA study, annotators determined a unified readability level (UL) for each sentence. The UL falls within the Max-Min range of annotator labels 99.2% of the time and matches one of the annotators 86.8% of the time. Table 7 compares the micro-average performance of annotators in Phase 2, using both pairwise comparisons and the comparison between the UL and the rounded average level (AL) of annotators’ choices. Table 7 also presents the results mapped to lower granularity levels (7, 5 and 3). We observe that overall, the AL-UL distance is smaller than the average pairwise distance among the annotators, and that its  $\pm 1$  Acc is much higher, which suggests the average (AL) is more often than not closer to UL than any pair of annotators are to each other. The comparison across granularity levels shows that although the absolute Distance decreases, its relative magnitude (compared to the label range) increases. As expected, both Acc and  $\pm 1$  Acc are higher with coarser level groupings. Appendix A.5 presents the results for each annotator against UL.

**Error analysis** To better understand annotator disagreement, we manually analyzed 100 randomly selected sentences with divergent readability labels. Table 8 presents representative examples

	19 Level	7 Level	5 Level	3 Level
<b>Pairwise Distance</b>	0.95	0.39	0.30	0.23
<i>Relative to Range</i>	5.0%	5.5%	6.0%	7.5%
<b>Acc</b>	61.1%	73.1%	75.2%	80.0%
<b><math>\pm 1</math> Acc</b>	74.4%	92.0%	95.0%	97.3%
<b>AL-UL Distance</b>	0.52	0.26	0.22	0.18
<i>Relative to Range</i>	2.7%	3.7%	4.4%	5.9%
<b>AL-UL Acc</b>	61.2%	75.5%	78.9%	82.9%
<b>AL-UL <math>\pm 1</math> Acc</b>	90.1%	98.5%	99.4%	99.5%

Table 7: Comparison of pairwise agreement micro averages across level granularities for all Phase 2 IAA sentences. UL = Unified Label; AL = Average Label.

with explanations. We found that 25% of disagreements were due to basic linguistic features (e.g., morphology, syntax, spelling), 12% involved emotional or symbolic content, 18% related to general advanced vocabulary, and 45% stemmed from domain-specific terminology in STEM, Humanities, or Social Sciences. This suggests that specialized vocabulary is the leading source of inconsistency, often due to differing expectations about what counts as general versus domain-specific language, and how specialization is defined. Some variation also stems from subjective views on what an *educated* Standard Arabic reader should know. In the future, we plan to develop readability lexicons to anchor our guidelines, building on efforts like the SAMER Lexicon (Al Khalil et al., 2020) and the Arabic Vocabulary Profile (Soliman and Familiar, 2024), but targeting 19 levels.

## 5.4 Automatic Readability Assessment

To establish a baseline for sentence-level readability classification, we fine-tune AraBERTv02 (Antoun et al., 2020) using the Transformers library (Wolf et al., 2019). Training is conducted on an NVIDIA V100 GPU for three epochs with a learning rate of  $5 \times 10^{-5}$ , a batch size of 64, and a cross-entropy loss function for multi-class classification across 19 levels. Table 9 presents the model’s learning curve. We evaluate performance using varying proportions of the training data:  $\frac{1}{8}$ ,  $\frac{1}{4}$ ,  $\frac{1}{2}$ , and the full dataset. As shown in the table, model performance improves consistently with larger training data. Compared to the Phase 2 IAA micro averages (Table 6), the model’s best Distance is 15.3% higher, and its best Accuracy is 5.3% absolute (8.7% relative) lower. However, the QWK is only marginally lower by just 0.8% absolute.

For a more extensive discussion of the automatic annotation results, see Elmadani et al. (2025).



Sentence (Arabic)	A1	A2	A3	A4	A5	UL	MM	Comments
أبي.. أبي.. <i>Dad .. Dad .. [lit. my father .. my father ..]</i>	2	2	2	3	3	3	1	First person singular pronoun is level 3.
احتضان الأم لهم. <i>The mother's embrace for them.</i>	9	12	5	5	5	5	7	Disagreement over احتضان 'embrace': standard or dialect aligned.
أشعر بالتعب والجوع.. <i>I feel tired and hungry..</i>	9	9	9	9	4	9	5	Vocabulary describing emotions (level 9).
يتم ضمان حيادية الإدارة بموجب القانون. <i>Administrative neutrality is guaranteed by law.</i>	12	12	12	14	12	12	2	Disagreement over حيادية 'neutrality': general advanced or specialized.

Table 8: Examples of Annotator Disagreements with Unified Levels (UL) and Max-Min Differences (MM)

Train	Distance	Acc <sup>19</sup>	±1 Acc <sup>19</sup>	QWK	Acc <sup>7</sup>	Acc <sup>5</sup>	Acc <sup>3</sup>
12.5%	1.35	45.0%	61.3%	77.2%	56.8%	63.0%	71.3%
25.0%	1.33	46.9%	63.0%	77.6%	58.8%	64.3%	72.3%
50.0%	1.16	52.4%	68.1%	80.7%	62.9%	67.6%	74.0%
100.0%	1.09	55.8%	69.4%	81.0%	64.9%	69.1%	74.7%

Table 9: Performance at different training data sizes across multiple evaluation metrics.

## 6 Conclusions and Future Work

This paper presented the annotation guidelines of the Balanced Arabic Readability Evaluation Corpus (**BAREC**), a large-scale, finely annotated dataset for assessing Arabic text readability across 19 levels. With over 69K sentences and 1 million words, it is, to our knowledge, the largest Arabic readability corpus, covering diverse genres, topics, and audiences. We report high inter-annotator agreement (QWK 81.8% in Phase 2) that ensures reliable annotations. Benchmark results across multiple classification granularities (19, 7, 5, and 3 levels) demonstrate both the difficulty and feasibility of automated Arabic readability prediction.

Looking ahead, we plan to expand the corpus by increasing its size and diversity to include more genres and topics. We also aim to add annotations for vocabulary leveling and syntactic treebanks to study the effect of vocabulary and syntax on readability. Future work will analyze readability variations across genres and topics. Additionally, we intend to integrate our tools into a system that assists children’s story writers in targeting specific reading levels.

The **BAREC** dataset, its annotation guidelines, and benchmark results, are publicly available to support future research and educational applications in Arabic readability assessment.<sup>1</sup>

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## Limitations

One notable limitation is the inherent subjectivity associated with readability assessment, which may introduce variability in annotation decisions despite our best efforts to maintain consistency. Additionally, the current version of the corpus may not fully capture the diverse linguistic landscape of the Arab world. Finally, while our methodology strives for inclusivity, there may be biases or gaps in the corpus due to factors such as selection bias in the source materials or limitations in the annotation process. We acknowledge that readability measures can be used with malicious intent to profile people; this is not our intention, and we discourage it.

## Ethics Statement

All data used in the corpus curation process are sourced responsibly and legally. The annotation process is conducted with transparency and fairness, with multiple annotators involved to mitigate biases and ensure reliability. All annotators are paid fair wages for their contribution. The corpus and associated guidelines are made openly accessible to promote transparency, reproducibility, and collaboration in Arabic language research.

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# A BAREC Annotation Guidelines Cheat Sheet and Annotation Interface

## A.1 Arabic Original

مستوى يارقي	صف	ACTFL	عدد كلمات	تهجئة وإملاء	تصريف واشتقاق	تركييب نحوية	مفردات	فكرة ومحتوى		
أ	روضة-1	مبتدئ أدنى	1	• كلمات من مقطع واحد أو مقطعين	• الفعل المضارع المفرد	• كلمة واحدة	• اسم جنس • اسم علم (متداول بسيط تركيبيا) • ضمير منفصل • مفردات متطابقة مع العامية - سامر I • الأرقام (العربية أو الهندية) 1-10	• فكرة مباشرة وصريحة وحسية. • لا رمزية في النص.		
							• جملة اسمية (هو يلعب) • إضافة حقيقية (باب البيت) • صفة وموصوف (باب كبير)	• فعل • صفة • مفردات متشابهة مع العامية - سامر I • العدد الأصلي بالأحرف • الأسماء الخمسة: أبوه، أخو		
							• سوابق: ال التعريف • سوابق: واو العطف • لواحق: ضمير المتكلم المفرد المتصل	• بدل كل: (صديقي أحمد) • بدل إشارة: (هذا البيت)	• مفردات فصيح شائعة - سامر I • اسم الإشارة المفرد • الأرقام (العربية أو الهندية) 100-11	
							• كلمات تستخدم مد الألف (إ)	• فعل المضارع الجمع • سوابق: حروف جر متصلة • ظرف منون	• جملة فعلية بدون مفعول به • جار ومجرور	• حروف الجر
ب	مبتدئ أدنى	≤2	• كلمات من 3 مقاطع	• كلمات من 4 مقاطع	• لواحق: ضمير متصل مفرد أو جمع • المثنى (في الأسماء والصفات) • جمع المؤنث السالم	• جملة فعلية مع مفعول به واحد اسم • جمل معطوفة • أدوات استفهام أساسية: ماذا، متى، من، أين، ما، كيف • صيغة التعجب "ما أفعل"	• العدد الترتيبي • الأرقام (العربية أو الهندية) 101-1,000 • اسم إشارة مثنى، جمع	• المحتوى من حياة القارئ. • لا رمزية في النص.		
							• كلمات من 5 مقاطع	• الفعل الماضي المفرد والجمع • جمع مذكر سالم	• جملة فيها فعلين (مثلا جملة فعلية مفعولها أن المصدرية)	• مفردات فصيح - سامر I
ج	1	مبتدئ متوسط	≤4	• كلمات من 6+ مقاطع	• فعل الماضي المثنى • فعل المضارع المثنى • فعل الأمر المفرد • لواحق: ضمير المثنى المتصل • جمع التذكير • واو القسم (والله)	• مفعول فيه (ظروف زمان ومكان) • حال • أداة الاستفهام هل	• مفردات فصيح شائعة - سامر II	• بعض الرمزية أو عدم التصريح المباشر بكل المقصود في الجملة		
د		مبتدئ متوسط	≤6	• كلمات من 11 ≤	• فعل الأمر الجمع • نون النسوة في الأسماء والأفعال • سوابق أخرى: سين الاستقبال، واو الاستئناف، فاء العطف • أدوات ربط (ثم، حي، أم، أمه، لكن، أما)	• المفعول المطلق • المفعول لأجله • المفعول معه • جملة فعلية تتعدى إلى مفعولين	• مفردات فصيح - سامر I و سامر II • أحرف الفتي • الأرقام (العربية أو الهندية) 1,001-1,000,000	• بعض الرمزية • يحتاج معها القارئ إلى مساعدة من يشرح له المقصود من الفكرة		
هـ		مبتدئ أعلى	≤8	• كلمات من 12 ≤	• فعل الأمر للمثنى • أداة الاستفهام: أ (أسمعت؟) • باء القسم • القسم: أداة القسم والمقسم به وجواب القسم.	• المنادى	• مفردات تصف حالات مزاجية وشعورية إيجابية وسلبية مثل الفرح، السعادة، الغضب، الأسف، الحسرة	• هناك شيء من الرمزية على مستوى الحدث في الجملة يتركها القارئ بنفسه أو من خلال معارفه السابقة		
و	2	مبتدئ أعلى	≤9	• كلمات من 15 ≤	• المبني للمجهول	• إن وأخواتها • كان وأخواتها • خبر مقدم / مبتدأ مؤخر • العنونة/السند • زب (حرف جر شبه بالتراند) • جملة الصلة وجملة الصفة • جملة الحال وجملة المفعول به	• أسماء الوصل المفردة • (قد - لقد) • (مما - عما - عى - علام - فيم - إلام - بـ...)	• مفردات فصيح - سامر I و سامر II • أحرف الفتي • الأرقام (العربية أو الهندية) < 1,000,000 • ذو • (زل - بلى - أجل - قط)	• هناك درجة من الرمزية وحاجة للمعرفة السابقة كي يفهم المقصود من الجملة	
ز		متوسط أدنى	≤10	• المشتقات العاملة (مثلا اسم الفاعل)	• التصغير	• جمل اعراضية (تفسير، دعاء) • استثناء • حصر • بدل (مثلا بدل بعض أو اشتمال) • تمييز	• مفردات فصيح - سامر III • اسم الفعل (مثلا أمين) • الأرقام (العربية أو الهندية) < 1,000,000 • ذو • (زل - بلى - أجل - قط)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• أفكار رمزية ومعنى باطن خاصة على صعيد البعد النفسي لل شخصيات أو الأحداث.	
ح		متوسط أدنى	≤11	• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها	
ط	3	متوسط أوسط	≤12		• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ي	4	متوسط أوسط	≤15		• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ك		متوسط أعلى	≤20		• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ل	5	متقدم أدنى			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
م	6-7	متقدم أوسط			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ن	8-9	متقدم أعلى			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
س	10-11	متقن أدنى			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ع	12	متقن أوسط			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ف	جامعة 2-1	متقن أعلى			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ص	جامعة 4-3	متقن			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
ق	متخصص	متميز			• نون التوكيد • ناء القسم	• الجمل شرطية ( مركبة - عادية) • حرف الجزم لما	• كلمات تصف حالات نفسية صعبة مثل الاكتئاب، الضياع، الاستغفار النفسي • استخدام كلمات منحوتة غير متداولة (مثلا هرجع • الخفيف الأحق مشتقة من هرجع و هجج) • الرموز (ش.م.)	• مفردات فصيح - سامر IV • مفردات قانونية، علمية، دينية، سياسية... غير متخصصة/عامه • (فو - حصر	• مفردات فصيح - سامر V • مفردات متخصصة ومفردات عربية عالية غير شائعة كثيرا في القضاء العام. • مفردات في الغالب بعيدة عن اللهجات العامية.	• أفكار رمزية، مجردة، علمية، أو شعرية وتحتاج إلى معارف لغوية ومعرفية سابقة للبناء عليها لأجل فهمها
هذا الـوسم يستخدم في حالة وجود صعوبة في تقييم المستوى، المفضل استخدام هذا الـوسم حتى تتمكن ك فريق عمل أن تجد حلا (مثلا بتعديل المعايير أو إضافة تفاصيل شريحة لها)										
هناك صعوبة في تقييم المستوى، المفضل استخدام هذا الـوسم حتى تتمكن ك فريق عمل أن تجد حلا (مثلا بتعديل المعايير أو إضافة تفاصيل شريحة لها)										
هناك مشكلة بصورة عامة، نستخدم هذا الـوسم للـجمل الحاسوبية على:										
• إخطاء إملائية (مثلا هزرات، تاء مربوطة، ألف مقصورة/ياء) • إخطاء في التشكيل • رككة لغوية (أمية، عامية، ترجمة سيئة من لغة أجنبية) • مواضيع غير لائقة (عنصرية، جنسية، تنمرية، إباحية، إلخ) • جمل وجارات معظمها مكتوب بلغات غير العربية أو بغير الخط العربي										
ولكن في الحالات التالية نوسم الجمل ونضيف أحد الحروف التالية في عمود الملاحظات: • خطأ في هزة الوصل/هزة << (أ) • كلمات خائفة << (ع) • إخطاء في التشكيل في بداية الجملة << (ت) • إخطاء غير المنقطعة في آخر الكلمة << (ي)										

هناك صعوبة هذا الوسم يستخدم في حالة وجود صعوبة في تقييم المستوى، يفضل استخدام هذا الوسم حتى تتمكن ك فريق عمل أن تجد حلا (مثلا بتعديل المعايير أو إضافة تفاصيل شرحية لها) هناك مشكلة بصورة عامة، نستخدم هذا الوسم للجمال (الحاوية على: أخطاء إملائية (مثلا همزات، تاء مريوطة، ألف مقصورة/ياء) أخطاء في التشكيل رككة لغوية (أمية، عامية، ترجمة سيئة من لغة أجنبية) مواضيع غير لائقة (عنصرية، حيادية، تمردية، إباحية، إلخ) جمل وعبارات معظمها مكتوب بلغات غير العربية أو بغير الخط العربي

ولكن في الحالات التالية نوسم الجمل ونضيف أحد الحروف التالية في عمود الملاحظات: خطأ في همزة الوصل/همزة القطع << (أ) كلمات خادشة << (ع) الخطأ في التشكيل في بداية الجملة << (ت) الياء غير المنطوقة في آخر الكلمة << (ي)



## A.2 English Translation

[illegible]

### A.3 Annotation Interface

Sentence/Phrase	Length	Level	Word Count	Spelling/Pronunciation	Morphology	Syntax	Vocabulary	Idea/Content	Notes
الجملة \ العبارة	عدد الكلمات	المستوى	عدد الكلمات	تهجئة/إملاء	تصريف واشتقاق	تركييب نحوية	مفردات	فكرة / محتوى	ملاحظات
خَيْرٌ	1	و (صف 2)	6-waw	٩ هو أعلى عدد كلمات مطبوعة غير متكررة بنون علامات الترقيم	• كلمات من ٥ مقاطع (بنون) • حساب حركات الإعراب	• الفعل الماضي المفرد والجمع • جمع منكر سالم	• جملة فيها فطين (مثلا) • جملة فعلية مفعولها أن (المصدرية)	• مفردات فصيحة - ١ سامر	• المحتوى من حياة القارئ. • لا رمزية في النص.
جودي يقربي	2	ز (صف 2)	7-zay	١٠ هو أعلى عدد كلمات مطبوعة غير متكررة بنون علامات الترقيم	• كلمات من ٦ مقاطع أو أكثر (بنون) • حساب حركات الإعراب • أفعال/أسماء معطلة الآخر	• الفعل الماضي المثني • الفعل المضارع المثني • فعل الأمر المفرد • جمع التذكير • وار القسم (والله)	• مفعول فيه (ظروف) • زمان ومكان • حال • أداة الاستفهام هل	• مفردات فصيحة شائعة - ٢ سامر	• بعض الرمزية أو عدم التصريح المباشر بكل المقصود في الجملة
بيروت في يوليو ١٩٦٦	4	ح (صف 3)	8-ha	١١ هو أعلى عدد كلمات مطبوعة غير متكررة بنون علامات الترقيم	• فعل الأمر الجمع • نون النسوة في الأسماء والأفعال (انتظرن) • دورهن • مبروك أخرى: سين الاستقبال، واو الاستئناف، فاء العطف • (ثم، حتى، أو، أم، لكن، أما)	• المفعول المطلق • المفعول لأجله • المفعول معه • جملة فعلية تتعدى إلى مفعولين	• مفردات فصيحة - ١ سامر • ٢ سامر • أحرف النفي • الأرقام (العربية أو الهندية) 1,000,000-1,001	• بعض الرمزية يحتاج معها القارئ إلى مساعدة من يشرح له المقصود من الفكرة	
كتابة خطة لمشروع الوحدة	4	ك (صف 4)	11-kaf	٢٠ هو أعلى عدد كلمات مطبوعة غير متكررة بنون علامات الترقيم	• المشتقات على أنواعها (تركز على المشتقات) • العلامة لاسميا اسم الفاعل واسم المفعول	• جملة اسمية خبرها • جملة اسمية (فيها مبتدآن) • إضافة خيالية (لفظية) • طويل القامة	• أسماء الوصل المثني والجمع • متلازمات لفظية مثل شارد الذهن، وارف الظلال	• هناك درجة من الرمزية وحاجة للمعرفة السابقة كي يفهم المقصود من الجملة	
اجتمع أهل في العيد	4	و (صف 2)	6-waw	٩ هو أعلى عدد كلمات مطبوعة غير متكررة بنون علامات الترقيم	• كلمات من ٥ مقاطع (بنون) • حساب حركات الإعراب	• الفعل الماضي المفرد والجمع • جمع منكر سالم	• جملة فيها فطين (مثلا) • جملة فعلية مفعولها أن (المصدرية)	• مفردات فصيحة - ١ سامر	• المحتوى من حياة القارئ. • لا رمزية في النص.
ولا يُخالطنا غمز ولا خور	4	ل (صف 5)	12-lam	لا حد لعدد الكلمات المطبوعة	• التصغير	• جمل اعتراضية (تفسير - دعاء...) • استثناء • اسم الفعل : إيه - صنة - امين - خي - هلام - هك - هيا - هيت - هلم إلى - مة - رويك • الأرقام (العربية أو الهندية) 1,000,000 < • تمييز • (يل - يلي - أجل)	• مفردات فصيحة - ٣ سامر • اسم الفعل : إيه - صنة - امين - خي - هلام - هك - هيا - هيت - هلم إلى - مة - رويك • الأرقام (العربية أو الهندية) 1,000,000 < • تمييز • (يل - يلي - أجل)	• هناك درجة من الرمزية وحاجة للمعرفة السابقة كي يفهم المقصود من الجملة	

This is a screenshot of the Google Sheet interface used for annotation. The first two columns on the left are the sentence and its word count. The third column is the readability level which is selected by drop down menus. The fourth yellow column and the first yellow row are not part of the interface, we added them for the purpose of explaining the structure to readers of this paper who do not know Arabic. The next 6 columns automatically display the text features from the annotation guidelines to help the annotators confirm their choices. The last column is for extra notes such as flagging problematic sentences.

### A.4 Annotation Team

	A0 <sup>P</sup>	A1	A2	A3	A4	A5 <sup>L</sup>
Native Language	Arabic	Arabic	Arabic	Arabic	Arabic	Arabic
Other Language	En, Fr	En	En, Fr	En, Fr	En, Fr	En, Fr
Nationality	Syrian	Lebanese	Lebanese	Lebanese	Lebanese	Lebanese
Residence	USA	Lebanon	Lebanon	Lebanon	UAE	Lebanon
Gender	Female	Female	Female	Female	Female	Female
Background	Muslim	Muslim	Muslim	Muslim	Christian	Muslim
Degree	MA	BA	BA	MA	MA	B MA
Major	Applied Ling.	Arabic Lit.	Geography	Arabic Lit.	Arabic Lit.	Arabic Lit.
Experience	CT, LA, RA	PT, LA	PT, LA	CT, LA	CT, LA	CT, LA, RA
School	Private	-	-	Public&Private	Private	Public
Level	University	Elementary	Elementary	Secondary	Secondary	Secondary
Students	L2	L1	L1	L1	L1	L1
Years	16	16	22	22	8	25

Table 10: Annotator background information. All have extensive linguistic annotation experience. Certified Teacher (CT), Private Tutor (PT), Linguistic Annotator (LA), Research Assistant (RA). A0<sup>P</sup> is the preprocessing and segmentation lead; and A5<sup>L</sup> is the readability annotation lead.

### A.5 Inter-Annotator Agreement between Annotator Labels and Unified Labels

	<b>Acc<sup>19</sup></b>	<b>±1 Acc<sup>19</sup></b>	<b>Dist</b>	<b>QWK</b>	<b>Acc<sup>7</sup></b>	<b>Acc<sup>5</sup></b>	<b>Acc<sup>3</sup></b>
<b>A1</b>	78.4%	89.0%	0.42	93.4%	85.3%	87.0%	89.7%
<b>A2</b>	65.1%	76.4%	0.87	82.2%	71.6%	73.6%	79.3%
<b>A3</b>	66.4%	78.4%	0.78	86.0%	73.7%	75.8%	79.0%
<b>A4</b>	63.7%	76.6%	0.86	83.8%	71.8%	74.2%	79.5%
<b>A5</b>	85.1%	91.2%	0.31	94.8%	89.2%	90.3%	92.9%
<b>Avg</b>	71.7%	82.3%	0.65	88.1%	78.4%	80.2%	84.1%

Table 11: Inter-Annotator Agreement (IAA) results comparing initial annotations by A1-A5 to unified labels (UL).

### B BAREC Corpus Level Distributions Across Splits

<b>Level</b>	<b>All</b>	<b>%</b>	<b>Train</b>	<b>%</b>	<b>Dev</b>	<b>%</b>	<b>Test</b>	<b>%</b>
<b>1-alif</b>	409	1%	333	1%	44	1%	32	0%
<b>2-ba</b>	437	1%	333	1%	68	1%	36	0%
<b>3-jim</b>	1,462	2%	1,139	2%	182	2%	141	2%
<b>4-dal</b>	751	1%	587	1%	78	1%	86	1%
<b>5-ha</b>	3,443	5%	2,646	5%	417	6%	380	5%
<b>6-waw</b>	1,534	2%	1,206	2%	189	3%	139	2%
<b>7-zay</b>	5,438	8%	4,152	8%	701	10%	585	8%
<b>8-ha</b>	5,683	8%	4,529	8%	613	8%	541	7%
<b>9-ta</b>	2,023	3%	1,597	3%	236	3%	190	3%
<b>10-ya</b>	9,763	14%	7,741	14%	1,012	14%	1,010	14%
<b>11-kaf</b>	4,914	7%	4,041	7%	409	6%	464	6%
<b>12-lam</b>	14,471	21%	11,318	21%	1,491	20%	1,662	23%
<b>13-mim</b>	4,039	6%	3,252	6%	349	5%	438	6%
<b>14-nun</b>	10,687	15%	8,573	16%	1,072	15%	1,042	14%
<b>15-sin</b>	2,547	4%	2,016	4%	258	4%	273	4%
<b>16-ayn</b>	1,141	2%	866	2%	114	2%	161	2%
<b>17-fa</b>	480	1%	364	1%	49	1%	67	1%
<b>18-sad</b>	103	0%	67	0%	13	0%	23	0%
<b>19-qaf</b>	116	0%	85	0%	15	0%	16	0%
<b>Total</b>	69,441	100%	54,845	100%	7,310	100%	7,286	100%

Table 12: Distribution of sentence counts and percentages across readability levels and data splits.

## C BAREC Corpus Sources

We present the corpus sources in groups of their general intended purpose.

Some datasets are chosen because they already have annotations available for other tasks. We list them independently of other collections they may be part of. For example, dependency treebank annotations exist (Habash et al., 2022) for the texts we included from the Arabian Nights, Quran and Hadith, Old and New Testament, Suspended Odes, and Sara (which comes from Hindawi Foundation).

### C.1 Education

**Emarati Curriculum** The first five units of the UAE curriculum textbooks for the 12 grades in three subjects: Arabic language, social studies, Islamic studies (Khalil et al., 2018).

**ArabicMMLU** 6,205 question and answer pairs from the ArabicMMLU benchmark dataset (Koto et al., 2024).

**Zayed Arabic-English Bilingual Undergraduate Corpus (ZAEBUC)** 100 student-written articles from the Zayed University Arabic-English Bilingual Undergraduate Corpus (Habash and Palfreyman, 2022).

**Arabic Learner Corpus (ALC)** 16 L2 articles from the Arabic Learner Corpus (Alfaifi, 2015).

**Basic Travel Expressions Corpus (BTEC)** 20 documents from the MSA translation of the Basic Traveling Expression Corpus (Eck and Hori, 2005; Takezawa et al., 2007; Bouamor et al., 2018).

**Collection of Children poems** Example of the included poems: My language sings (لغتي تغني), and Poetry and news (أشعار وأخبار) (Al-Safadi, 2005; Taha-Thomure, 2007).

**ChatGPT** To add more children’s materials, we ask Chatgpt to generate 200 sentences ranging from 2 to 4 words per sentence, 150 sentences ranging from 5 to 7 words per sentence and 100 sentences ranging from 8 to 10 words per sentence.<sup>6</sup> Not all sentences generated by ChatGPT were correct. We discarded some sentences that were flagged by the annotators. Table 13 shows the prompts and the percentage of discarded sentences for each prompt.

<sup>6</sup><https://chatgpt.com/>

### C.2 Literature

**Hindawi** A subset of 264 books extracted from the Hindawi Foundation website across different different genres.<sup>7</sup>

**Kalima** The first 500 words of 62 books from Kalima project.<sup>8</sup>

**Green Library** 58 manually typed books from the Green Library.<sup>9</sup>

**Arabian Nights** The openings and endings of the opening narrative and the first eight nights from the Arabian Nights (Unknown, 12th century). We extracted the text from an online forum.<sup>10</sup>

**Hayy ibn Yaqdhan** A subset of the philosophical novel and allegorical tale written by Ibn Tufail (Tufail, 1150). We extracted the text from the Hindawi Foundation website.<sup>11</sup>

**Sara** The first 1000 words of *Sara*, a novel by Al-Akkad first published in 1938 (Al-Akkad, 1938). We extracted the text from the Hindawi Foundation website.<sup>12</sup>

**The Suspended Odes (Odes)** The ten most celebrated poems from Pre-Islamic Arabia (المعلقات Mu’allaqat). All texts were extracted from Wikipedia.<sup>13</sup>

### C.3 Media

**Majed** 10 manually typed editions of Majed magazine for children from 1983 to 2019.<sup>14</sup>

**ReadMe++** The Arabic split of the ReadMe++ dataset (Naous et al., 2024).

**Spacetoons Songs** The opening songs of 53 animated children series from Spacetoons channel.

**Subtitles** A subset of the Arabic side of the Open-Subtitles dataset (Lison and Tiedemann, 2016).

**WikiNews** 62 Arabic WikiNews articles covering politics, economics, health, science and technology, sports, arts, and culture (Abdelali et al., 2016).

<sup>7</sup><https://www.hindawi.org/books/categories/>

<sup>8</sup><https://alc.ae/publications/kalima/>

<sup>9</sup>[https://archive.org/details/201409\\_201409](https://archive.org/details/201409_201409)

<sup>10</sup><http://al-nada.eb2a.com/1000lela&lela/>

<sup>11</sup><https://www.hindawi.org/books/90463596/>

<sup>12</sup><https://www.hindawi.org/books/72707304/>

<sup>13</sup><https://ar.wikipedia.org/wiki/المعلقات>

<sup>14</sup>[https://archive.org/details/majid\\_magazine](https://archive.org/details/majid_magazine)

Prompt	Targeted #Words per Sentence	Prompt Text	% Discarded
Prompt 1	2-4	I am creating a children's textbook to practice reading in Arabic. I need short sentences containing 2 to 4 words that are limited to children's vocabulary. Give me 200 sentences in Standard Arabic -- no need to include English.	1.5%
	Examples	الشمس مشرقة. البنات تأكل الفاكهة.	
Prompt 2	5-7	I am creating a children's textbook to practice reading in Arabic. I need 5-word, 6-word, and 7-word sentences that are limited to children's vocabulary. Give me 150 sentences in Standard Arabic -- no need to include English.	1.3%
	Examples	الأسد ينام تحت شجرة كبيرة. الأطفال يلعبون في الملعب ويضحكون بسعادة كبيرة.	
Prompt 3	8-10	I am creating a children's textbook to practice reading in Arabic. I need long sentences (8-word, 9-word, and 10-word sentences) that are limited to children's vocabulary. Give me 100 sentences in Standard Arabic -- no need to include English.	1.0%
	Examples	الأرنب يقفز فوق العشب الأخضر في الصباح الباكر. الغرد يتسلق الأشجار بسرعة ويقفز ببراعة من فرع إلى فرع.	

Table 13: ChatGPT Prompts. % Discarded is the percentage of discarded sentences due to grammatical errors.

## C.4 References

**Wikipedia** A subset of 168 Arabic wikipedia articles covering Culture, Figures, Geography, History, Mathematics, Sciences, Society, Philosophy, Religions and Technologies.<sup>15</sup>

**Constitutions** The first 2000 words of the Arabic constitutions from 16 Arabic speaking countries, collected from MCWC dataset (El-Haj and Ezzini, 2024).

**UN** The Arabic translation of the Universal Declaration of Human Rights.<sup>16</sup>

## C.5 Religion

**Old Testament** The first 20 chapters of the Book of Genesis (Smith and Van Dyck, 1865).<sup>17</sup>

**New Testament** The first 16 chapters of the Book of Matthew (Smith and Van Dyck, 1860).<sup>17</sup>

**Quran** The first three Surahs and the last 14 Surahs from the Holy Quran. We selected the text from the Quran Corpus Project (Dukes et al., 2013).<sup>18</sup>

**Hadith** The first 75 Hadiths from Sahih Bukhari (al Bukhari, 846). We selected the text from the LK Hadith Corpus<sup>19</sup> (Altammami et al., 2019).

<sup>15</sup><https://ar.wikipedia.org/>

<sup>16</sup><https://www.un.org/ar/about-us/universal-declaration-of-human-rights>

<sup>17</sup><https://www.arabicbible.com/>

<sup>18</sup><https://corpus.quran.com/>

<sup>19</sup><https://github.com/ShathaTm/LK-Hadith-Corpus>