

# ArBanking77: Intent Detection Neural Model and a New Dataset in Modern and Dialectical Arabic

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## Contributions

## ArBanking77 dataset

- ArBanking77 dataset consists of 31,404 queries.
- 2.4x larger than the Banking77 dataset.
- On average, there are 408 queries per intent
  - 202 MSA queries/intent
  - 206 Palestinian queries/intent.

## BERT Model

• F1-scores on MSA and PAL are 0.9209 and 0.8995, respectively

Used AraBBERT-V2 pre-trained model (Antoun et al., 2020)

## ArBanking77 Dataset

## The original Banking77 dataset

- 13,083 queries
- 77 classes (intents)
- Single domain, banking
- Open under the (CC-BY-4.0) license

### The ArBanking77 corpus:

Each query in the original Banking77 has at least two corresponding queries in the ArBanking77

- At least one query written in MSA.
- At least one query written in Palestinian dialect.

#### **Annotation Process**

### **Annotation Phases:**

**Phase 1:** Arabization and Localization

**Step1:** The translation of the Banking77 from English into MSA.

Done using Google Translate API.

**Step2:** The manual annotation .

The annotators performed four steps for each original English query:

- I. MSA\_1 should be revised in case of incorrect translation.
- II. MSA\_2 is optionally written by the annotator.
- III. PAL\_1 is the formulation of the query in the Palestinian dialect.

#### Phase 2: Review

**Step1:** Each annotator reviewed 3 belonged intents, to ensure that:

- I. The MSA and Palestinian queries should be acceptable, semantically correct and well-formulated.
- II. All queries in one intent belong to that intent, and not to other intents (labeling consistency).
- III. Spelling mistakes are ignored in order to simulate common errors and noise in real NLP systems, especially in live chat queries.

**Step2:** We revised duplicate queries by introducing additional variations to make them unique.

### **Lexical Relation between MSA and PAL**

- 26 annotators (Well trained)
- Done using Google Sheets
- Over several months

IV. PAL\_2 is optionally written by the annotator.

Each intent was divided among 2-5 annotators.

 Measured using the Jaccard Index for each parallel pair (MSA and PAL)

#### Results of Jaccard index:

The mean is 0.16, the median 0.13, and the standard deviation 0.13.

## **Intent Detection Model**

- BERT encoder is fine-tuned on Arabic intent detection task using the ArBanking77 dataset.
- A single linear layer was added on top of BERT transformer layers to perform the intent classification task.

## **Experiments and Results**

- Zero-Shot Cross-Lingual Transfer Learning
  Used multi-lingual BERT (mBERT) (Devlinetal.,2018) and GigaBERT (Lanetal.,2020).
- Low-Resource Simulation
  The size of samples one needs to achieve good and acceptable accuracy.

#### **Results:**

	N	ISA Test		P	AL Test	
% of data	Precision	Recall	<b>F1</b>	Precision	Recall	<b>F1</b>
20%	0.8825	0.8755	0.8758	0.8441	0.8403	0.8363
50%	0.9117	0.9094	0.9088	0.8909	0.8903	0.8888
100%	0.9231	0.9212	0.9209	0.9004	0.9025	0.8995

## **Downloads and Demo**

https://sina.birzeit.edu/arbanking77/

Public (data, code, demo)



#### **Result:**

Pre-trained Model	Training Data	MSA F1	PAL F1
Multi-lingual BERT (uncased)	ArBanking77 (MSA)	-	0.5968
GigaBERT	Banking77 (English)	0.5047	0.3507
Multi-lingual BERT (uncased)	Banking77 (English)	0.1774	0.0903

#### • Pre-Trained Transformers Benchmark

Evaluate various Arabic pre-trained transformer models, we benchmark against these models:

#### **Results:**

	MSA Test			PAL Test				
<b>Pre-trained Model</b>	Precision	Recall	<b>F</b> 1	Precision	Recall	<b>F1</b>		
AraBERTv2	0.9231	0.9212	0.9209	0.9004	0.9025	0.8995		
MARBERTv2	0.9161	0.9142	0.9138	0.8983	0.8981	0.8962		
ARBERT	0.9103	0.9121	0.9115	0.8810	0.8923	0.8899		
QARiB	0.9147	0.9123	0.9121	0.8846	0.8864	0.8835		
CAMeLBERT-Mix	0.9149	0.9133	0.9128	0.8855	0.8854	0.8830		
MARBERT	0.9106	0.9075	0.9070	0.8817	0.8817	0.8789		
Multi-lingual BERT	0.8888	0.8872	0.8862	0.8598	0.8623	0.8578		

#### Noise and Error Simulation

Experimented with three types of error and noise simulations:

common spelling errors simc
 simulated errors sims

3. keyboard-related errors simk

#### Results

Train Augmentation	Test Augmentation	MSA Test PAL T		PAL Test			
	Test Augmentation	20%	50%	100%	20%	50%	<b>J0%</b> .8995 J.8637 0.8463 0.863 0.901
	None	0.8758	0.9088	0.9209	0.8363	0.8888	0.8995
None	$sim_c$	0.8452	0.8795	0.8981	0.7933	0.8435	0.8637
	$sim_s$	0.8454	0.8813	0.8893	0.7585	0.8269	0.8463
	$sim_k$	0.8392	0.8648	0.8844	0.7942	0.8428	0.8634
	None	0.8801	0.9126	0.9207	0.8421	0.8901	0.9018
$sim_s/sim_k$	$sim_c$	0.8583	0.8922	0.9001	0.8065	0.8602	0.8711
	$sim_s$	0.8683	0.9017	0.9121	0.8055	0.8641	0.8857
	$sim_k$	0.8499	0.8833	0.8909	0.8086	0.8529	0.8749



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ArBanking77						
A dataset and source-code for ArBanking77 Version: 1.0 (updated on 1/9/2023)						
ArBanking77 consists of 31,404 (MSA and Palestinian dialect) that a card arrival, card linking, exchange rate, and automatic top-up. A	re manually Arabized and localized from the original English Banking77 dataset; which consists of 13,083 queries. neural model based on AraBERT was fine-tuned on the ArBanking77 dataset (F1-score 92% for MSA, 90% for PAL). T	. Each query is classified into o Try the service (type sentences	ne of the 77 c seperated by	lasses (i newLine	ntents) i e or ° or	ncluding ? or ! or .
		لم تسللي بطاقتي أين هي؟				
<u>IN</u>	Detect					
	وصول البطاقة					
	لم تسىلانى بىلاتكى أين هى؟					
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