A Linguistic Analysis of the Problems of Facebook (see translation) and Google (translate) Applications: A Selected Sample of Corpora دراسة لغوية تحليلية لمشكلات الترجمة الآلية لتطبيقي فيسبوك وجوجل لعينة من المتون Dr. Heba Abdelraheim Ibrahim Alkady Lecturer in Linguistics & Translation Studies Faculty of Arts - South Valley University د هبة عبدالرحيم إبراهيم القاضي مدرس اللغويات والترجمة كلية الأداب – جامعة جنوب الوادي

### A Linguistic Analysis of the Problems of Facebook (see translation) and Google (translate) Applications: A Selected Sample of Corpora

### Abstract

Machine Translation or (MT) is considered one of the recent innovations in technology in the field of translation studies (TS). This paper is an attempt to redefine nowadays users' methods of translating corpora from Arabic into English and vice versa. MT is a computational activity in which the translating process is done by using a bilingual or multilingual data set. Corpus analysis is one of the fastest-growing methodologies in contemporary linguistics. Many institutions and individuals use MT to translate corpora. They resort to computational translation applications offered by Facebook (see translation) and Google (translate) to render their corpora (words, utterances, statements, speech, texts...etc. Regardless of the fatal mistakes sometimes found in the results of MT (For example, those mistakes related to diacritical markers), no one can deny that MT systems are built on gigantic lexical banks and dictionaries. This paper presents a linguistic analysis of the problems of MT namely Facebook (see translation) and Google (translate) applied to a limited sample of corpora.

### Keywords:

Machine Translation, Diacritical Markers, Facebook (see translation), Google (translate), Corpora.

## دراسة لغوية تحليلية لمشكلات الترجمة الآلية لتطبيقي فيسبوك وجوجل لعينه من المتون

#### ملخص البحث

تعتبر الترجمة الآلية إحدى صور الإبتكار في الصناعات اللغوية الحاسوبية وتمثل تقنية مذهلة في مجال دراسات وعلم الترجمة. يمثل هذا البحث محاولة التعرف أكثر علي وسائل وتطبيقات مُستحدثه شائعة بين مستخدمي تطبيقات الترجمة العربية والانجليزية بمنظور آلي جديد وتحديد المشكلات التي تواجه هذا النمط الآلي عن طريق تحليل لغوي لعينة من المتون المفردات والعبارات والمنشورات الفيسبوكية التي تمثل ظواهر وصعوبات لغوية متنوعة من المفردات ورغم نقائص وتحمي تطبيقات الترجمة العربية والانجليزية بمنظور آلي جديد وتحديد المشكلات التي تواجه هذا النمط الآلي عن طريق تحليل لغوي لعينة من المتون المفردات والعبارات والمنشورات الفيسبوكية التي تمثل ظواهر وصعوبات لغوية متنوعة من ورغم نقائص وعيوب الترجمة الآلية عن طريق نظام معالجة ثنائي اللغة. ورغم نقائص وعيوب الترجمة الآلية التي تعود للسياق اللغوي والخصائص التركيبية النحوية والصرفية والالالية للمفردات التي تمثل صعوبة لتطبيقات الترجمة الآلية او علمات التركيبية النحوية التربية والعمان وعيوب الترجمة الآلية عن طريق نظام معالجة ثنائي اللغة. ورغم نقائص وعيوب الترجمة الآلية التي تعود للسياق اللغوي والخصائص التركيبية النحوية والصرفية والاشتقاقية والدلالية للمفردات التي تعود للسياق اللغوي المعران والمؤلية التركيبية النحوية المعرفي والخصائص التركيبية النحوية المرغم نقائص وعيوب الترجمة الآلية التي تعود للسياق اللغوي والخصائص التركيبية النحوية والصرفية والالالية للمفردات التي تمثل صعوبة لتطبيقات الترجمة الآلية اوعلامات التشكيل الا أنها مرجع ووسيلة سريعة يستخدمها كثير من الأفراد والمؤسسات في ترجمة المفردات والنصوص من وإلي الإنجليزية.حيث لوحظ ازدياد إقبال المستخدمين علي تطبيقات المفردات والنصوص من وإلي الإنجليزية.حيث لوحظ ازدياد والم المشردات والمؤسات في ترجمة الراسات في ترجمة القردات والنو مثلي في ترجمة المفردات المفردات والمؤمية والإلية مرفي وجوجل.حيث لوحظ ازدياد ولاي الممردات والمؤسات في ترجمة المفردات والنصوص من وإلي الإنجليزية.حيث لوحظ ازدياد ولابال المستخدمين علي نيرجمة المفردات والنو مثل فيسبوك وجوجل.تهدف الدراسة إلي تقديم تحليل لغوي لمام المشكلات المفردات والتي قيسبوك وحوجل.تهدف الدراسة إلي تعديم ولي المفردان المشكلات من المفردان والم مألي فيسبوك وحوجل.حيف المرالما ولي ولو المشكاني المشكلات المفريات المشكلية

**الكلمات الدالة:** الترجمة الآلية, علامات التشكيل, تطبيق ترجمة فيسبوك, تطبيق ترجمة جوجل, المتون.

### A Linguistic Analysis of the Problems of Facebook (see translation) and Google (translate) Applications: A Selected Sample of Corpora

### Introduction

Arabic Language is spoken by a great ratio of the world's population. It is a powerful means of social control. It is used as a means of communication among Arabs and non-Arabs as well. Accordingly, Arabic is becoming a crucial language on the Internet due to the increasing number of Arabic speaking online users (*Facebookers & Googlers*) seeking Arabic content and translation applications, so it cannot be sidelined.

The rapid advance of online services to satisfy users is important to enhance the Internet usage. Google *(translate)* and Facebook *(see translation)* applications are examples of this advancement. These applications involve the capability to support multiple languages and provide tools to offer and use multi-lingual content.

The key of the corpora used in this study is authentic language. The corpus approach (Biber, Conrad & Reppen, 1998:4) is used because it is empirical, analysing the actual patterns of language use. The corpora used in this study are composed of written text, a sample of Facebook public and personal posts, class lectures I give to my students as users of machine translation applications and a random sample of corpora and online texts. The main purpose is to determine how the various linguistic patterns and usage of corpora (a particular word, sound, text, syntactic, or morphological constructions) result in inaccurate machine translation.

## Google (translate) and Facebook (see translation)

Computer technology has been applied in technical translation in order to improve speed and cost of translation (Trujillo, 1999). Translation with the aid of machines can be faster than manual translation and can reduce the cost of translation. In addition, the use of machine translation (MT) can result in improvements in quality, particularly in the use of consistent terminology within a scientific text or for a specific domain Google (*translate*) and Facebook (*see translation*) are the examples of the most currently used multilingual interne-based machine translation applications. The growing influence of the Internet has been reflected in the appearance MT applications like Google and Facebook investments in the area of Machine Translation to enhance their services to the Arab world users. In April 2006, Google (*translate*) was launched and originally created by Franz Josef Och with a statistical machine translation engine (Franz, 2006).

Google (*translate*) does not apply grammatical rules, since its algorithms are based on statistical analysis rather than traditional rulebased analysis. Google (*translate*) does not translate from one language into another (L1 to L2). Instead, it often translates first into English and then into the target language (L1, En, L2) (Christian et al, 2016). Some languages produce better results than others. Google (*translate*) supports 103 languages. English to and from Arabic was launched in April 28, 2006.

Facebook, a basically social networking service, was founded 13 years ago by Mark Zuckerberg. Its service covers the whole world except three blocking countries. Facebook's latest update does not provide the service (see translation button). The following figures show how these applications work:

Figure 1: (Facebook "see translation" App)

How do I get the Translate Facebook app?					
in ne	nteer translators help make Facebook available w languages, and can help improve the lations for existing languages.				
Face Face Face	ne who wants to bring their language to book or improve the current translations on book can be a translator. People who use book in your language can see your approved lations.				
Get	the Translate Facebook App				
To ge	et the translate Facebook App:				
1	From the main page of your app, tap <b>Search</b> at the top.				
2	Type in "translate facebook" and tap <b>Search</b> .				
З	Tap <b>Apps &gt; Translate Facebook &gt; Open</b> .				
4	Select the language you want to translate into. We recommend using Facebook in the same language that you're translating				

- -

Go	Google Translate homepage					
Type of site	Machine translation					
Available in	103 languages, see below					
Owner	Google					
Website	translate.google.com &					
Commercial	Yes					
Registration	Optional					
Users	Over 200+ million people daily					
Launched	April 28, 2006; 10 years ago (as statistical machine translation) <sup>[1]</sup> November 15, 2016; 3 months ago (as neural machine translation) <sup>[2]</sup>					
Current status	Active					

Figure 2 (Google "Translate" Homepage)

It is apparent that translation plays a substantial role in human communication. There have been many translating strategies and types of equivalence. Some scholars prefer word- for- word translation, whereas others select sentence for sentence translation. A translator carries the burdens of conveying the meaning and effect of the text from one language to another.

To facilitate this task, several websites offering automatic / MT services, which translate not only corpus, sentences or even long document, were launched. Google Translation Service (GST) offers, state-of-the-art free translation service and works automatically without the intervention of human translators.

For several languages, one may see a speaker button near the translated text and by clicking this icon; one can hear a machinegenerated spoken version of the translation. The Google translator allows translating whole documents, for example, in the form of PDF, TXT, DOC, PPT, XLS or RTF, or even images by just clicking the "translate a document" link and submitting a file without the need for copying and pasting large blocks of text (Sternby et.al., 2009). A machine translation (MT) system is actually based on descriptions of both the source language corpus and the target language corpus at all levels: algorithm, formal grammars and vocabularies to produce translations. This machine translation process is based on the following steps (Eynde, 2015: 4):

- □ Analyze source language text based on vocabulary, morphological and syntactical analysis
- □ Conversion (translation of source text to target text)
- $\Box$  Synthesis creation of text for target language based on syntactical and morphological appearance of text.

All these steps in machine translation system may be interrelated and/or absent.

### **Literature Review**

Several studies have been conducted in the field of Arabic Language used online. Much research started with studies focusing on Arabic data input, character set analyzers, identification systems, search engines, machine translation and Arabic content online. Many of these efforts had contributed significantly to the field. Among these unique studies are the works of Beesley (1998)"Arabic Morphological Analysis on the Internet", Sternby, Morwing, Andersson and Friberg (2009) "On-Line Arabic handwriting recognition with templates", Selamat and Ng (2011) "Arabic script web page language identifications using decision tree neural networks," Shen and Khalifa (2009)"Facebook usage among Arabic college students: preliminary findings on gender differences", Zantout and Guessoum (2001)"An Automatic English-Arabic HTML page translation system", Sanan, Rammal, and Zreik, "Internet Arabic search engines studies". There have been plentiful efforts studying the effect of applying morphological processing (Hui, 1998; Habash and Sadat, 2006; Oflazer and Durgar El-Kahlout, 2007; Badr et al., 2008). Error analysis of MT using an open-source tool for error analysis of natural language processing tasks targeting morphologically rich languages was conducted by El Kholy and Habash, 2011. Habash and others have interesting works on Arabic Morphological Representations in Machine Translation (2004, 2005, 2006). The general history of MT is covered and updated by by Hutchins (1986, 1988, 1994, 1999, and 2001).

For systems developed during the 1990s the main sources are the survey of techniques by Trujillo (1999), the biennial "MT Summit" conferences and the numerous annual conferences and workshops for MT (e.g. Farwell et al. 1998, White 2000), computational linguistics (principally the Coling and ACL conferences), artificial intelligence, and information retrieval. In addition to other studies devoted to Arabic to English Machine Translation by Salem, Hensman and B. Nolan (2008).

The area of applying machine translation analysis to applications like <u>Google translate</u> and <u>Facebook see translation</u> is still a virgin territory.

### **MACHINE TRANSLATION (MT) of Arabic Corpora**

With the late technological advancements in MT, Arabic has received attention in order to automate Arabic translations (Farghaly et al., 2009). The accuracy of human translation is very far from MT, as MTA does not guarantee that punctuation and spelling are one hundred percent precise. The corpus provided here is a sample which is frequently used and given to students as an example of simple sentences with complex and various layers of meaning and word orders. Google and Facebook are chosen because they provide freely available translations.

Therefore, there are issues and errors in MT, resulting in confusion regarding (word order) who is doing what or to whom or who reports to whom etc., see the following examples:

Arabic has a variety in word orders; this is a very urging problem in MTA due to the infinite prospects to express one sentence in Arabic.

1. V + N + N 2. N + V + N

This means that the subject and the object of each sentence have to be precisely identified. Table1 shows this challenge:

	SL (Arabic)	Gloss		MT		TL (English)
	ي ُحب محمد	Yu: I	Hib	He	loves	Mohamed
1	علي	muhamed 'ali		Mohamed Ali		Loves Ali
يُحب علي محمد 2	Yu: Hib	fali	I love	Ali	Mohamed	
	يحب علي محمد	muhamed		Mohame	ed	Loves Ali

#### Table 1

The Difference between both examples is in the position of the actor. In example (1) the actor is the first argument of the verb.

In example (2) the actor is the second argument of the verb. Both sentences have the same meaning.

#### **Study Questions**

Through MT of a sample of Arabic and English variety of lexis, phrases and texts, the study tries to answer the following questions:

- a. How efficient/deficient is MTA parallel corpora (target language output)?
- b. What are the linguistic, grammatical, morphological, and contextual patterns associated with MTA inaccuracies?
- c. What are the common errors of MTS?
- d. How does MTS application functionally work?

#### **Sample of Corpus Analysis**

When translating Arabic, Machine Translation Applications have some semantic challenges, lexical choice mismatches or lexical and structural ambiguities. My corpora are presented according to prior set requirements, criteria with the purpose of showing the various linguistic patterns of Arabic corpora that pose a problem in MT. They include (lexis, phrases, concordances, and long sentences or utterances) See the following examples:

The screenshots below show how (*Google translate application*) or *GTA* translated the Arabic corpus 'aqd into English. The Arabic  $(n)^{\frac{1}{2}}$  in English would be translated in this context as necklace.

In screenshot 1, the SL عِقَدُ is typed without adding diacritics or the definite article ال and the result is inappropriate TL equivalent.



In screenshot 1, the SL عقد is typed without adding any inflectional diacritics and the result was inappropriate TL equivalent.

#### Screenshot (2):



In screenshot 2, the SL as a a a a b a critised the result was inappropriate TL equivalent.

# Screenshot (3): Arabic - detected ← ب ب ب ب ب اعطیتها العقد کی <u>تلبسه</u> aietituha aleaqd kay talbisuh Did you mean العقد کی تلبسه

### I gave her contract to wear

In screenshot 3, Inflectional diacritics "?lif laam " (U) of definiteness are added to the SL and still we have the same misplaced TL equivalent.

### **Corpus Source**

This paper purports to analyze the linguistic properties of a sample of SL corpora that result in inaccurate machine translation and improper TL parallel corpora. The Corpora are taken from some of the assignments I give to my university students during translation classes and various publically and personally shared Facebook posts. This papers shows how MT users, Facebook (*see translation*) and Google (*translate*) users apply machine translation to produce a parallel corpora. The sample of analysis does not provide imaginary idealized examples.

### Aims and Methodology

This paper aims at tackling some of the linguistic problems facing the better utilization of Arabic Language on Internet based Machine Translation Applications (MTA). This paper presents a linguistic analysis of some of the problems resulting from the use of applications of MT, namely Facebook (*see translation*) and Google (*translate*). The paper is applied to a selected sample of corpora comprised of formal and informal Arabic lexis, texts and Facebook posts. The sample is randomly selected to show the various linguistic patterns of Arabic and English languages. The corpora are presented in the form of screenshots taken from the MTAs webpages. All screenshots of the corpora collected are shown in the Appendix. The paper also discusses the challenges facing the Arabic language on internet based MTA; namely, Google (*translate*) and Facebook (*see translation*). In this paper, the impact of both non or partial diacritization on machine translation (MT).

### The Challenges of Arabic Corpora to MT

An intact translation is one from which the TL parallel corpora readers can ably recognize the correct meaning of the SL corpus or text. Recent Machine Translation Application systems still make many blatant inaccuracies of meaning. Here are some of the issues of MTA to Arabic:

- 1. The problem of finding equivalent TL corpus for many language pairs especially when the SL is morphologically opulent.
- 2. A Singular Feminine adjective is mostly formed only by suffixing "taa? marbuta" to the masculine adjective. For example "mudarrisa" is the feminine of "mudarris" (teacher). Notice the following screenshots of GTA English rendering of the Arabic word مُدَرِسة in both examples:



Screenshot (4)

Screenshot (5)

In screenshot (4), the source text is a single word (root+ feminine ending) with no diacritics.

In screenshot (5), the same word occurs in a clear context. The source text is a complete sentece ( V+S+O). In both cases , the result is an inaccurate translation.

3. Lexical diacritics differentiate between two lexemes. For example, the diacritization difference between the lexemes (بَاحِتْ) bæHlø 'researcher' and (بَاحَتْ) bAHaøa 'negotiated' distinguishes between the meanings of the word (lexical ambiguation) rather than their inflections.

The following screenshots show how GTA produced an imprecise translation:



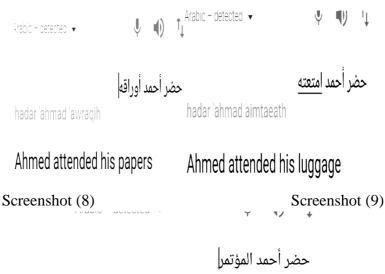
Besides, Having a non-diacritically marked Arabic text results in lexical and morphological ambiguity. Clear examples of this problem can be found on Facebook pages like "*Insan*" "أينسان" "in which there are multiple mistranslated Arabic quotes :

تشتد و تشتد ثم يأتي الفرج

ta<sup>f</sup>ttad wa ta<sup>f</sup>ttad oumma ya?ti alfarra3

The word farra3 is mistranslated as "vulva" instead of "relief" or "ease" which is a fatal mistake in translation and a sharp violation of the meaning of the sentence

4. The *shadda* (gemination) diacritic adds further meanings to the lexeme خَضَرَ *HaDar* 'attend' versus خَضَرَ *HaDDar* 'prepared'. Notice the following three screenshots (8, 9, and 10) that show these English rendering of the two verbs by GTA.



hadar 'ahmad almutamar

# Ahmed attended the conference

Screenshot (10)

5. Arabic has no copula verb "to be".

In Arabic there is no copula verb "to be" [Abn-Aqeal, 2007]. The verb to be' is understood and then predicate subject. The subject and the predicate have to be in the nominative case, an example is shown in the table below:

Table 2

Table 2						
Arabic	Gloss	Google Translation	English			
محمد طالِب	Mohamed Tælib	Mohammed Taleb	Mohammed			
محمد صابِب		Wonannieu Taleo	(is) a student			
محمد مُعلِم	Mohamed	Mohamed Maalim	Mohamed (is)			
	Mu'alim	Wonameu Waamm	a teacher			
محمد عَالِم	Mohamed	Mohamed world	Mohamed (is)			
	'a'ælim	Wonamed wond	a scientist			
محمد َطْبِيب	Mohamed T <sup>3</sup> bi:b	Mohammed	Mohamed is a			
		doctor	doctor			

- 6. The Arabic Language has various attachable clitics which include conjunction proclitics, e.g. y w 'and', particle proclitics, e.g., J l+ 'to/for', the definite article Language Al 'the', and the class of pronominal enclitics, e.g., hum 'their/them'.
- 7. Arabic has the specific case of dual, whereas other languages move From the singular to the plural form directly. In Arabic, we need only to add two letters to the singularform to express the dual form.

An example is given in table (3):

Table	3
	-

Table 3			
Source Language	Gloss	Google Translation	Target Language
باب	Bæb	Door	Door
بابين	Bæbain	Babin	Two doors
بابان	Babæn	Babæn	Two doors
جرسين	zarasayn	jursæn	Two bells
جرسان	zarasæn	Jrsan	Two bells
امرأتين	i'mra?atain	Two women	Two women
قلم	q°l°m	Pen	Pen
قلمين	Qalamain	Two pencils	Two pens- pencils
بقرة	Baqara	A cow	A cow
بقرتين	Baqaratain	Two cows	Two cows
بقرتان	Baqarta:n	Bqrtan	Two cows
ز هر ة	Zahra	A flower	flower
ز هرتان	Zahrata:n	Flowers	Two flowers
ز هر تین	Zahratain	z-hrtyn	Two flowers
شجرتين	∫aʒarain	Shjrtyn	Two trees
شجرتان	∫aʒara:n	Wood	Two trees

9. The variety of the Arabic corpora used by *Facebookers* and *Googlers* does not include diacritics. However, diacritical markers are extremely

useful for clear readability and comprehension. Having a nondiacritically marked Arabic corpora results in lexical and morphological ambiguity.

10. Arabic diacritics are vowelization marks and usually absent (Schlippe, et al 2008:271). "*shadda*" is the only diacritic which appears in several modern Arabic scripts (Ibid). Native speakers distinguish the right pronunciation and the correct meaning of a word without diacritic marks by considering the context and the position of the word in a sentence.

11. There are four types of spelling errors:

- 1) word separation errors; 2) misspellings;
- 3) wrong capitalization; 4) wrong punctuation.

#### Analysis

The following table of analysis is based on a comparison of MT and human Translation of a limited sample of random and frequently shared corpora associated with many Facebook pages and electronically stored corpus. Screenshots of the following corpora including the full electronic text of each corpus and its MT "parallel corpus" are provided in the appendix.

Table (4) shows the source language corpora and their machine translation parallel corpora as provided by Google (*translate*) or Facebook (*see translation*) in addition to a gloss for the corpora. A human translation of the corpora is also provided in the table. The corpora chosen here serve to answer the question of this paper:

What are the linguistic, grammatical, morphological, and contextual patterns associated with MTA inaccuracies?

Answers of this question are shown in the analysis and notes column (Table 4).

Table 4
I dole 1

SL Corpora	Gloss	MT Parallel Corpora	TL Parallel Corpora	Analysis & Notes
يروي قلبي	Yarwi: qalibi	Tells	(lit.) quench my thirst or warm my heart	MTA is unable to predict context sensitive word connotation This is a concordance corpus whose meaning is not literal. It means warmed my heart MTA Lacks recognition of the concordance corpus meaning
لتمحو ها	litamHu:ha	To tamhouhha	To be abolished, rubbed out, erased	MTAproducesaTransliterationnottranslationThis is a morphologicallycomplex corpusMTAisunabletorecognizetheSLinflectionMultipleinflectionsadded to the stem maHa:
لا تبتلينا	La tabta:li:na	Tptlina	Don't let us be sorely tired by – afflicted with	MTA produced a Transliteration rather than translation MTA is unable to recognize negative forms MTA is unable to recognize multiple inflections
هيحاسبنا	haiHasibna	Hahasa- pena	Judge, consider	The corpora are examples of informal Arabic forms MTA is unable to recognize Arabic
بتطلع	bitTala'	Aspira-tion	Bring out	varieties Misplacement of lexical

ممکن	mumkin	possible	May	Equivalent
تجيبلي	Tigi:bli	tjabla	bring me	MTA is unable to disambiguate different meanings
مبتلبسوش	Mabtilbi:- su: ∫	mbtalp-suc	You don't wear	MTA cannot predict Vowelization
أوجاعكم	?aw3aa'a- kum	auajjae-km	Pains suffering	MTA cannot recognize multiple inflections MTA produced a
بسجدة	bisa3da	bsgda	With a prostration	Transliteration not a translation
ستصيبك	satusi:b∧k	stsepk	Bless with	
تحسب	taHsab	calculat-ed	Consider- think	
ماشيه	ma ∫ia	cattle	believe	
بجد	bi3ad	hard	real	
ب <u>جد</u> مش	meſ	Mesh	Not	This corpus is an informal negative form
عشان	ʻa∫an	Ashan	because, for	MTA is unable to
وتشيله	W t <sup>√</sup> i:luh	and chilh	and or then Keep store	recognize informal Arabic or language variations
تاني	ta:ni	tani	Again	
کي تري احلام	ila 'alHulm	الي الحلم	To dream	MTA lacks recognition or prediction of the SL corpus syntactic category
Fair	ʻaadil	عادل	جميل - ناعم	and verb forms.

				MTA is unable to recognize multiple semantic connotations of a corpus MTA is unable to
with u	Ma'a ju:	u مع	معك	recognize new corpus forms ((abbreviations and typed characters)
Let down	Tadaa' ?asfal	تدع أسفل	تخذل	MTA cannot predict the semantic reference of certain corpora concordances (phrasal verbs)
				MTA is unable to recognize the function of lexicogrammar (relating the preposition to the verb)
Spending	qaDa:a?	قضاء	أمضي- اقضي And the	MTA is unable to locate word's syntactic category and function in a certain context
فشارب	fa <sup>∫</sup> aarib <sup>∫</sup> aarib	Fsharb	drinker Drinker- drinking	MTA is unable to recognize the SL corpus <b>inflections</b> and complex <b>morphology</b>
شارب	Baaban	Sharp	Door Basem	Multiple <b>inflections</b> added to the stem <sup>f</sup> ariba are not recognized by
بابا	Basem Morsi (Name)	Baba	Morsi (Name of a person)	MTA MTA is unable to recognize <b>diacritics</b> "tanweeen"
باسم مرسي		Morsi Spokesman		MTA cannot recognize context and <b>syntactic</b> <b>category</b>

### **Conclusion and Results**

Through this limited sample of corpus of the MT of Arabic and English corpora provided in the study in form of screenshot or tables of analysis, it becomes very clear that there is a serious open problem related to MTAs recognizing grammatical, morphological, syntactic and contextual patterns of the corpora (as noted and explained in table 4).

- 1. Machine Translation applications are not robust across different types of data, performing poorly on corpora whose underlying properties differ from those of the translation application data.
- 2. Most MT systems inaccuracies occur when dealing with morphologically rich Arabic corpora. The MT application will not generate word forms that they have not observed.
- 3. The corpora represent various linguistic patterns which generate inadequate and unacceptable MTA renderings. The corpora prove that the translation generated by GTA and FSTA is inadequate and unsatisfactory form linguistic point of view. These improper generated results give a negative impression of these online applications.
- 4. The transliteration forms provided by *Google* and *Facebook* are not accurate.
- 5. Undoubtedly the various flaws of GTA and FSTA shown in the table (4) prove that they are far from replacing professional translators.
- 6. Users of MT should use their common sense, which may resolve translation ambiguities.
- 7. These translation services may provide us with database of parallel corpora or root words.
- 8. Arabic dual word form is mostly unrecognized in GTA and FSTA.
- 9. These translation services should focus on providing multiple parallel corpora to polysemous word input that has multiple possibilities of translation.
- 10. These translation services should update their input methods database, and add diacritical markers and context sensitive recognition systems for the Arabic language.
- 11. Morphology reveals a number of translational problems, not all of which are the same type as shown in the table of analysis.
- 12. Machine translation that does not process language's spirit can only turn put a corpse of parallel corpus.

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### Appendix

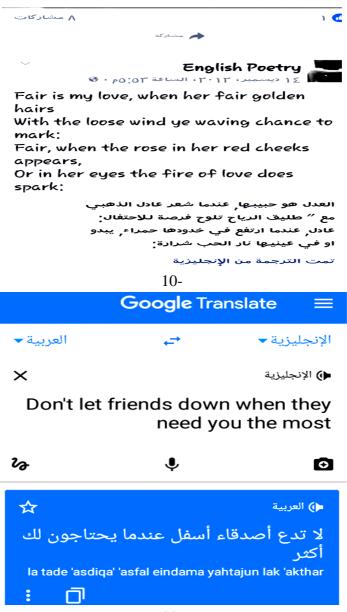
Screenshots of the Corpora Collected

1-

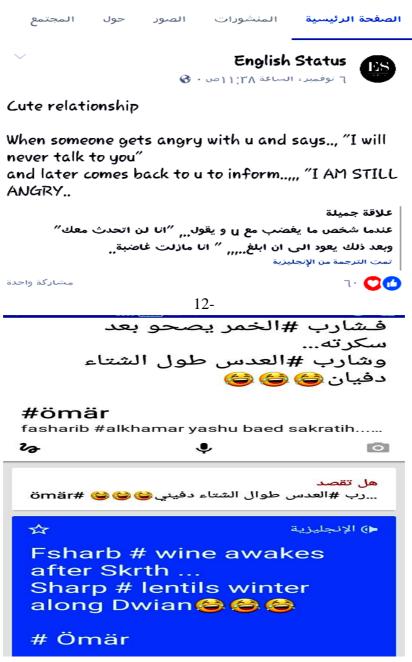








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