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English-Persian Phrase-based Statistical Machine Translation: Enhanced Models, Search and Training

A THESIS PRESENTED IN FULFILMENT OF THE REQUIREMENTS FOR THE DEGREE OF

DOCTOR OF PHILOSOPHY
IN
COMPUTER ENGINEERING

at Massey University Albany (Auckland), New Zealand

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ABSTRACT

Machine translation (MT), as applied to natural language processing, has undergone substantial development over the past sixty years. While there are a number of different approaches to MT, there has been increasing interest in statistical machine translation (SMT) as the preferred approach to MT. Advances in computational power, together with the exploration of new methods and algorithms have enabled a general improvement in the output quality in a number of systems for various language pairs using this approach. However, there is a significant lack of research work in the area of English/Persian SMT, mainly due to the scarcity of data for this language pair, and the shortage of fundamental resources such as large-scale bilingual corpora. Several research studies have been published on work in the area of machine translation involving the Persian language; however, results producing fluent, usable output are rare.

This thesis shows how SMT was implemented with this language pair for the first time, and how we created a cutting-edge hybrid SMT system capable of delivering high-quality translation output.

We present the development of what is currently the largest English/Persian parallel corpus, constructed using a web crawler to source usable online data, together with the concatenation of existing parallel corpora. As yet another contribution of the research, we propose an improved hybrid corpus alignment method involving sentence length-based and word correspondence-based models to align words, phrases and sentences in the corpus. We also show the impact that the corpus domain can have on the translation output, and the necessity to consider domains of both bilingual and monolingual corpora where they are included in the training and language models.

Two open-source toolkits, Moses and Joshua, were modified to work with the Persian language, and their behaviour and performance results were compared to determine which performed better when implemented with the Persian language.

We present our work in designing, testing, and implementing a novel, three-level Transfer-based automatic post-editing (APE) component based on grammatical rules, which operates by analysing, parsing, and POS-tagging the output, and implements functions as transformers which perform corrections to the text, from lexical

transformation to complex syntactical rearrangement. We show that rule-based approaches to the task of post-editing are superior to the commonly-used statistical models, since they incorporate linguistic knowledge, and are strong in terms of syntax, morphology, and structural semantics – qualities which are very desirable when performing grammatical correction and syntactical restructuring.

We implement independent manual evaluation as well as standard automatic techniques, in order to assess more accurately the translation output. This evaluation shows that the use of the APE component is able to improve translation output significantly, that is, by at least 25%, resulting in high-quality translation output.

Our system performs well by using a combination of the capabilities of two main MT approaches – SMT and RBMT – in different areas of the system as a whole. SMT provides the main system with consistent, mathematical-based translation, and the Transfer-based algorithm in the APE component operates with comprehensive linguistic rules in order to improve incorrect sentences, and fine-tune translation output. This results in a robust, state-of-the-art system, which noticeably exceeds other currently available solutions for this language pair.

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In the name of God who owns soul and wisdom. These are the best attributes of God. ~ Ferdowsi (935 – 1020)

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Dedicated to my parents

DECLARATION

The author declares that this is her own work except where due acknowledgement has been given. It is being submitted for a PhD in Engineering to Massey University, New Zealand.

This Thesis describes the research carried out by the author at the School of Engineering, Massey University, New Zealand, from June 2008 to December 2012, supervised by Dr Rezaul Hasan.

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LIST OF ABBREVIATIONS

ACL Complement Clause of Adjective

ADV Adverb

ADVC Adverbial Complement of Verb AJCONJ Conjunction of Adjective

AJPP Prepositional Complement of Adjective

AJUCL Adjunct Clause AOL America Online

APE Automatic Post Editing
APOSTMOD Adjective Post-Modifier

APP Apposition

APREMOD Adjective Pre-Modifier

ASR Automatic Speech Recognition

AVCONJ Conjunction of Adverb

BLEU Bilingual Evaluation Understudy
CAT Computer Assisted Translation
CFGs Synchronous Context-Free Grammars

CNW Canada Newswire COMPPP Comparative Preposition

C-STAR International Consortium for Research on Speech Translation

DARPA Defence Advance Research Project Agency

DG Dependency Grammar

EBMT Example-Based Machine Translation

EGIU English Grammar in Use

ELRA European Language Resource Association EM Expectation–Maximization Algorithm

En English

ENC Enclitic Non-Verbal Element

Fa Farsi

FAMT Fully Automatic Machine Translation

FLDB Farsi Linguistic Database FST Finite State Transducer

FTD US Air Force's Foreign Technology Division

GLP Gross Language Product

HAMT Human-Assisted Machine Translation

Hiero Hierarchical

HMT Hybrid Machine Translation

IBM International Business Machines Corporation
IEEE Institute of Electrical and Electronics Engineers, Inc.

IR Information Retrieval

IRNA Iranian News Agency

IWSLT International Workshop on Spoken Language Translation

LG Link Grammar
LV Linking Verb
LVP Light Verb Particle

MAHT Machine-Assisted Human Translation

MAP Maximum A-Posteriori

MERT Minimum Error Rate Training

MESU Measure

METEOR Metric for Evaluation of Translation with Explicit Ordering

MLE Maximum Likelihood Estimation

MOS Mosnad

MOZ Ezafe Dependent

MPEC Modern Persian-English corpus

MST Parser Maximum Spanning Tree Parser

MT Machine Translation
NADV Adverb of Noun
NCL Clause of Noun
NCONJ Conjunction of Noun

NE Non-Verbal Element of Infinitive NEZ Ezafe Complement of Adjective

NIST National Institute of Standards and Technology

NLP Natural Language Processing
NPEC News Persian English Corpus
NPOSTMOD Post-Modifier of Noun
NPP Preposition 0f Noun
NPREMOD Pre-Modifier of Noun
NPRT Particle of Infinitive

NSPEC News Subtitle Persian-English Corpus

NVE Non-Verbal Element

ODJ Object

ODJ2 Second Object OOV Out of Vocabulary

PAHO Pan American Health Organization

PARCL Participle Clause
PART Interrogative Particle

PB Phrase Based

PCONJ Conjunction of Preposition PCTS Parallel Corpus Test Set

PeEn-SMT Persian-English Statistical Machine Translation

PEN Parallel English-Persian News Corpus

POS Part of Speech
POSDEP Post-Dependent
PPL Perplexity Threshold
PREDEP Pre-Dependent
PREM Pre-Modifier
PRO Predicate

PROG Progressive Auxiliary
PUNC Punctuation Mark

RBMT Rule-Based Machine Translation

RHS Right Hand Side

ROOT Root

SAMT Syntax Augmented Machine Translation

SBJ Subject

SCFG Stochastic Context-Free Grammar SDL Scalable Enterprise Translation Server

SDL Language Weaver

SMT Statistical Machine Translation

SOV Subject-Object-Verb SRILM Sri Language Model SVO Subject-Verb-Object

TAM Tamiz

TEP Tehran English-Persian Corpus
TER Translation Error Rate Te
TER Translation Error Rate
TMC Tehran Monolingual Corpus

U Unicode

UN **United Nations**

University of Tehran Information Retrieval Evaluation System Complement Clause of Verb UTIRE

VCL

VCONJ

Conjunction of Verb
Prepositional Complement of Verb VPP

Verb Particle VPRT WER Word Error Rate

WSD Word Sense Disambiguation

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