

Machine Translation of Controlled Languages for More Reliable Human Communication in Safety Critical Applications¹

1. Introduction

The research results presented in this paper concern linguistics and controlled language machine translation, an application within the LiSe (Linguistique et Sécurité) project² [1], this in the context of security and crises in general, and in particular where communication ought to be rapid and correct [2].

2. Controlled Language Machine Translation

In an emergency or crisis, not only have alert messages to be controlled at source [3, 4], but we do not necessarily know the messages that must be translated beforehand. As well as this, the translated messages too have to be controlled for the target language audiences. However any machine translation must be accomplished with neither manual pre- nor post-edition as both are unacceptable in an emergency due to time constraints. In the LiSe project, controlled French has been the source controlled language (SCL) and for the target controlled languages (TCLs) these are controlled French (identity – exploited for testing), Arabic, Chinese, English (abbreviation An) and Thai; see Figs. 1 and 2.

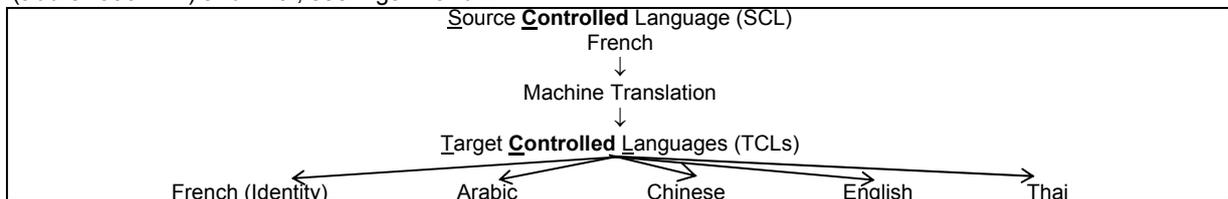


Figure 1. Machine translation of controlled languages

Corpus source: http://www.interieur.gouv.fr/misill/sections/a_votre_service/votre_securite/conseils-incendie/incendie-chez-vous/view

UncontrolledText	SCL	ControlledText	SCL = TCL?	TCL	Translation
Ne branchez pas trop d'appareils sur la même prise	French	Ne pas brancher trop d'appareils sur la même prise	SCL = TCL	French	Ne pas brancher trop d'appareils sur la même prise
			SCL ≠ TCL	Arabic	رايتلا بيش نم سن فن بن قزمج ال انم ري ثكفل اول صوت ال
				Chinese	不要把太多的电器插在同一个插销
				English	Do not connect too many electrical appliances to the same plug
Thai	อย่าเสียบปลั๊กเครื่องใช้ไฟฟ้าจำนวนมากเกินไปบนที่เสียบอันเดียวกัน				

Figure 2. Example illustrating Translation

3 Machine Translation Architecture

In our architecture, what is new is that both the source language SL (French) and the target language(s) TL(s) are controlled, not only to conform to normal controlled language constraints, but also mutually for translation knowing that each and all influence the others [5]. Pre-edition is avoided as well, as control of the source is provided during message entry by means of the 'SL to CPSL' User Interface [1]. The TLs being controlled, this obviates post-edition. All this has resulted in a novel hybrid (pivot + transfer) rule-based machine translation architecture in which the pivot language PL is French controlled also for translation (**C**ontrolled **P**ivot **S**ource **L**anguage CPSL), and where the Transfer System is directed by the various source-target language divergences. In the case of French, we have French → French: identity (∅ divergences). To situate our architecture and its scalability, Fig. 3 illustrates how our own "Controlled Pivot Source Language + Transfer" architecture is related to conventional pivot and transfer architectures.

¹ Keynote speech in this Symposium.

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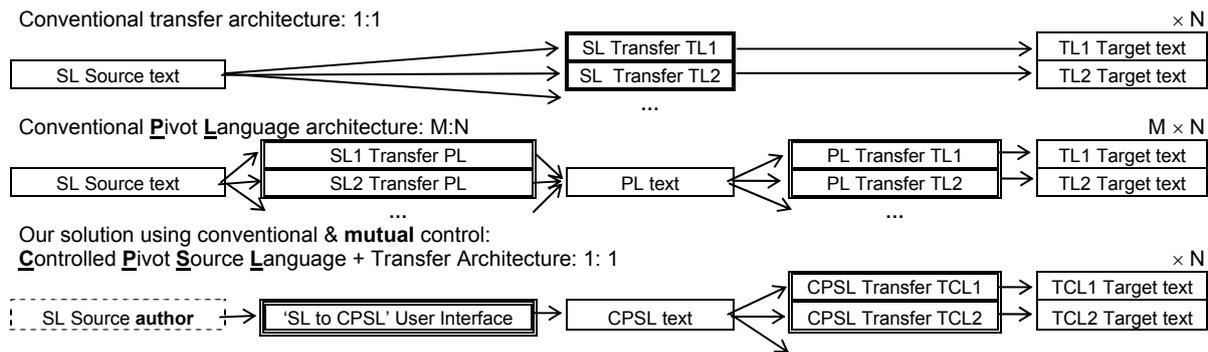


Figure 3. MT architectures, scalability and applicability

4. Author, Pivot Language & Transfer

The author composes messages syntagmatically in the controlled French pivot language using the 'SL to CPSL' User Interface. For example "Ne pas brancher trop d'appareils électriques sur la même prise." is expressed in the CPSL by the structure:

```
RegroupeageEnSyntagms_LS = [
    ['Ne',          neg1],
    ['pas',        neg2],
    [brancher,     vinf],
    ['trop d'appareils électriques', arg1],
    [sur,          prep_v],
    ['la même prise', arg2],
    [.,           pt]]
```

This structure is passed to the Transfer System for translation into each target language and the mutual divergences with the controlled French direct the translation process. Transfer divergences can occur at variously the sentence, syntagmatic and lexical levels. For example for the first two (these include the lexical level):

– Sentence Level

French:

```
opt(neg1) + opt(neg2) + vinf + arg1 + prep_v + arg2 + pt
Ne pas brancher trop d'appareils électriques sur la même prise.
```

Chinese:

```
opt(arg0) + opt(neg1) + indicateur(['把','让']) + arg1 + v + arg2
不要把太多的电器插在同一个插销
```

– Syntagmatic Level

```
ad + adj + n = adj + cl + n
`la même prise` = `同一个插销`
```

- The French *ad* has no corresponding lexis in Chinese.
- The Chinese *cl* has no corresponding lexis in French.
- The French *adj* corresponds to the Chinese *adj* (linked via the *dictionnaireLexical*).
- The French *n* corresponds to the Chinese *n* (linked via the *dictionnaireLexical*).

Transfer is effectuated for each source-target language couple as follows:

- Source language sentence level selection is determined by the 'SL to CPSL' User Interface which subsequently enables each unique corresponding target language sentence level structure to be selected (= in the case of French).
- Lexical analysis of each syntagm in the controlled French source sentence (the suite of syntagms delivered by the 'SL to CPSL' User Interface) is performed. This results not only in the construction of the lexical transfer table for each SCL/TCL couple but also the selection of the relevant syntagm transfer structure couple.
- Transfer is effectuated by the target language as requestor and **not** the source language as provider. This ensures that no linguistic rule placement coding is necessary where:
 - A source language (French) source lexis has no corresponding lexis in the target language
 - A target language lexis has no corresponding lexis in the French source.

Finally target surface level post transfer operations are effected: e.g. syntagm lexies (government) agreement (e.g. Arabic adjectives with nouns) and clitics.

5. Micro-Systemic Linguistic Analysis of Machine Translation

We now turn to the micro-systemic linguistic analysis that we have carried out [6, 7]. For each (\forall) linguistic phenomenon observed when translating from the Source Controlled Language to each Target Controlled Language we observe (\exists):

- a single 'canonical' case where there are no language divergences (identical source and target controlled language phenomena)

- 'variant' cases encompassing the divergences between each target controlled language and the source controlled language

The whole language identity case (variant = canonical) applies to all linguistic phenomena observed when the Source Controlled Language \equiv Target Controlled Language. For an observed linguistic phenomenon when translating, in classifying the variant cases, the linguist establishes two categorisations in the form of two partitions and then puts these into relation, one with the other. The categorisations are:

- 'non-contextual' (nc) categorisation of the canonical forms **C** in relation with the variant forms **V** in isolation, the context being limited to just the canonical and variant forms themselves. This results in the Partition P_{nc} .
- 'in-context' (ic) categorisation of the canonical forms in relation with the variant forms in terms of the linguistic contexts of the variant forms. The systemic analysis reveals precisely what other related linguistic systems are involved. This results in the Partition P_{ic} .

Given that we have partitions (P_{nc} and P_{ic}), from the fundamental theorem on equivalence relations, it follows that there exist two corresponding equivalence relations E_{nc} and E_{ic} , both over the binary ordered relation between the canonical forms and the variant forms **CV**. We model the system over the linguistic phenomenon by means of the binary ordered relation **Ss** between the equivalence relations E_{nc} and E_{ic} , each over **CV**. From this relation **Ss** we can generate algorithms.

6. Micro-systemic Translation Architecture

Applying micro-systemic linguistic analysis, we have classified and organised the equivalences and divergences in the form of a compositional micro-system structure. In this structure the resulting micro-systems are expressed in a declarative manner by means of typed container data structures in the form of a database together with their contents so as to be incorporated in the machine translation process; see Fig. 4 [8].

Super μ System	Explanation
Ss_TACT	Root μ system – T raduction A utomatique C entre T esnière
Ss_frC	Controlled French
Ss_frC_frC	Controlled French \rightarrow Controlled French (Identity – no divergences , constructed from Ss_frC)
Ss_frC_arC	Controlled French \rightarrow Controlled Arabic
...	...

Figure 4. Machine Translation Super Micro-Systems

In Figs. 5 and 6 we show the typed container data structures corresponding to **Ss_frC_frC** and **Ss_frC_arC**; note the presence of example & corpus attributes in the data base. For **Ss_frC_arC** we mark with a grey background the divergences with **Ss_frC_frC** (attribute and/or content); for example, unlike French, Arabic has a case system.

Ss_frC_frC:

table					
frC					
frC_groupesVerbaux_frC					
frC_args_frC					
frC_groupes					
frC_dictionnaireLexical_frC					
frC_catégories_frC					
frC_catégories					
frC_groupesVerbaux_frC					
frC_groupesVerbaux_frC	verbe_frC	verbe_frC	groupe_frC	exemple	corpus
frC_args_frC	arg_frC		arg_frC	exemple	corpus
frC_groupes	groupe_frC		structure_frC	exemple	corpus
frC_dictionnaireLexical_frC	lexique_frC	catégorie_frC_frC	lexique_frC	catégorie_frC	exemple corpus
frC_catégories_frC	catégorie_frC_frC		exemple	corpus	
frC_catégories	categoryie_frC		exemple	corpus	

Figure 5. Typed container data structure corresponding to **Ss_frC_frC**

Ss_frC_arC:

Table							
arC							
arC_groupesVerbaux_frC							
...							
arC_groupesVerbaux_frC							
arC_groupesVerbaux_frC	verbe_arC	verbe_arC	groupe_arC	exemple	corpus		
arC_args_frC	arg_frC		arg_arC	exemple	corpus		
arC_groupes	groupe_arC		structure_arC	exemple	corpus		
arC_dictionnaireLexical_frC	lexique_frC	catégorie_frC_arC	lexique_arC	noLexique_arC	accLexique_arC	...	tdefLexique_arC catégorie_arC...
arC_catégories_frC	catégorie_frC_arC		exemple		corpus		
arC_catégories	categoryie_arC		exemple		corpus		

Figure 6. Typed container data structure corresponding to **Ss_frC_arC**

7. Micro-systemic Translation Programming

For ergonomic linguistic programming reasons, the concrete (content containing) form of the typed container data structures is realised by means of spread-sheets upon which is mapped a typed & interpretable data structure

representing the Machine Translation micro-system. For example at the spread-sheet level: worksheet & column names' languages (other than for Ss_TACT) are formulated as follows (in BNF):

LS = Langue Source (SCL), LC = Langue Cible (TCL)
Full worksheet name :: LS | LC | LC_worksheetname | LC_worksheetname_LS
Full column name :: columnname | columnname_LS | columnname_LC | columnname_LS_LC

(Worksheet names & column names cannot contain '_'). At the cell level we have:

cell_types([atom, plus_to_list, term, atom_list]).

Fig. 7 summarises the target controlled language divergences with controlled French.

Sentential syntagmatic order	Transfer lexis POS ambiguity	Target clitic:
Syntagm lexical order		prefixing
Category (Part of Speech)		postfixing
Super Categories (POS, Case...)	Lexis:	infixing
Syntagm lexies agreement	Source	
Case (declension)	yes	yes
Classifier	yes	no
Indicator	no	yes

Figure 7. Target Language Divergence with Controlled French

In Fig. 8 we give examples of coding for some of these divergences.

Target Language Divergence with Controlled French	Controlled Language					Example coding
	frC	anC	arC	chC	thC	
Category (Part of Speech)	∅	yes	yes	yes	yes	frC: même adjs , arC: même det
Super Categories (POS, Case...)	∅	yes	yes	yes	yes	arC: n = [nms, nmp, nfs, nfp, nmph, nmpnh, nfph, nfpnh] arC: cases = [no, acc, t, nodef, accdef]
Transfer lexis POS ambiguity	∅	yes	yes	yes	yes	thC: frC: adj1 + nmp + adj2 thC: n + adj2 + adj1
Lexis						
Source	Target					
no	yes	∅	yes	yes	yes	arC: arC: artu_ + n_A(acc)+lexis_(‘ل’) + adj_A(acc)

Remark: the Lexis example also illustrates Case (declension) and Target clitics (artu_ and lexis_) – both prefixing clitics.

Figure 8. Examples of coding for some of the controlled language divergences with controlled French

Our implementation involves a single target language independent kernel in which the Target Language divergences with Controlled French are abstracted as propositions. For example: ‘Case (declension)’:

Database (spread-sheet cell content):

arC_catégories: cases = [no, acc, t, nodef, accdef]

anC_catégories: cases = []

Kernel code:

Cases_TL = [] ⇒ Target Language with case system (e.g. LC = arC Arabic)

Cases_TL = [] ⇒ Target Language without case system (e.g. LC = anC English)

8. Problems put into evidence and solved by our Controlled Language Machine Translation system

We take the case of the machine translation of French → Arabic [9] with an example from the domain of aeronautics:

Couper la pompe avant du réservoir de milieu d’aile droite.

This sentence, in French, is entered into the ‘SL to CPSL’ User Interface. From the target language buttons, if French is selected, as the divergences = ∅, one obtains the Identity ‘translation’ – this via the Transfer System:



Selecting the Arabic button results in:



- In this example one notes the system’s capacity for translating sequences of Nominal Groups nested with the preposition ‘de’. This type of sequence of NGs requires an in-depth analysis in respect of translation to Arabic for several reasons.
- There are two types of determination in Arabic; one being with the definite article and the other being a particular determination involving annexation.
- As well as this, Arabic has a morpho-syntactic characteristic linked to its inflexional morphology and to its case marking which concerns nouns and adjectives (Case (declension)).
- Syntactic linking has also to be taken into account because it is necessary to make the agreement between nouns and adjectives (Syntagm lexies agreement).
- To these problems must be added the cliticisation of certain parts of speech in Arabic (Target clitic).

9. Tracing

The controlled language machine translation system includes logged dynamic tracing:

- i. Input JSON (interface language (www.json.org) with the 'SL to CPSL' User Interface) text logged in pretty print format for subsequent automated benchmark/regression testing
- ii. For each target language:
 - a. Prepare for transfer :
 - A. RegroupageEnSyntagms_LS = value
 - B. LC_GroupeVerbal_LS = value
 - C. Regroupage_En_Arguments_LS = value
 - D. Unites_Source = value
 - E. Do transfer (driven by Target Language)
 - b. Regroupage_En_Arguments_LC = value
 - c. Post transfer:
 - A. government_agreement (if applicable)
 - B. Unites_Cible = value
- iii. Traductions = value
- iv. JSON to return to the 'SL to CPSL' User Interface - Translations

10. Implementation Process

To situate and relate our project in terms of the real world of application engineering we terminate with the implementation process devised in conjunction with the different project actors and which is summarized in Fig. 9.

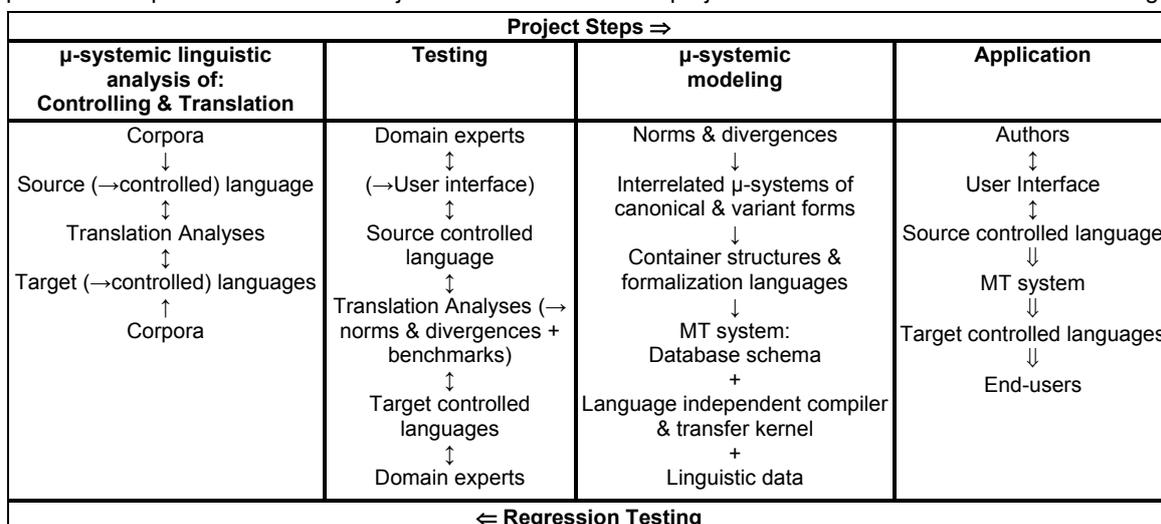


Figure 9. Controlled language Machine Translation implementation process

11. Conclusion

We have explained our novel hybrid rule-based machine translation architecture which involves as pivot language the source language controlled also for translation, and in which the transfer system encompasses source-target language divergences. We have related this to the real world of application engineering by terminating with the implementation process. We stress that the micro-systemic architectures for the controlled language machine translation have been devised by linguists, and that micro-systemic linguistic analyses are exhaustive and provide traceability, essential qualities in safety critical applications. In like manner the linguistic programming depends on concepts, formalisations and specialised coding languages conceived by linguists which together result in table driven declarative programming. Language is innately human and could well play the key role in the management of crises.

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