Building MT systems in low resourced EU languages for Public Sector users in Croatia, Iceland, Ireland and Norway

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[Re-]Introducing Language Weaver



Language Weaver. The last mile in machine translation.



Introducing The PRINCIPLE Project

- A 2-year project funded by the Connecting Europe Facility (CEF)
- Focused on collecting data to improve translation quality in the EU Digital Services Infrastructures (DSIs) for prioritised low-resourced EU languages.
- The main aim of the project is to identify, collect and process high-quality Language Resources (LRs) for the following under-resourced European languages:
 - Croatian

Icelandic

- Irish
- Norwegian (Bokmål and Nynorsk)



PRINCIPLE: The Role of Machine Translation

By building state-of-the-art Neural MT models with data collected in the PRINCIPLE project, two key objectives can be accomplished:



Benchmarking and evaluation of MT systems built using project data attests to the quality of data collected and its value for MT systems developed in Europe.



Granting free access and use of MT systems to Public Sector bodies during the course of the project provides an incentive for contributions of language data.

• Public sector bodies who participate in this incentive are labelled 'Early Adopters' in the PRINCIPLE project.



What Data Already Existed for These Languages?

Iconic completed a full search/download of existing resources from <u>ELRC-Share*</u>.

A quality review was conducted by PRINCIPLE project partners.

Language	# Resources	# Translation Units
Irish	41	901,421
Croatian	36	3,891,799
Icelandic	17	801,283
Norwegian	47	1,964,961
Norwegian (Nynorsk)	4	6,358





* https://elrc-share.eu/



What Data Already Existed for These Languages?

Iconic completed a full search/download of existing resources from ELRC-Share.

A quality review was conducted by PRINCIPLE project partners.

Data was then cleaned/filtered for MT Baseline system development.

Language	# Resources	# Translation Units	# TU used in MT Baseline
Irish	41	901,421	588,663
Croatian	36	3,891,799	3,337,608
Icelandic	17	801,283	702,139
Norwegian	47	1,964,961	1,140,351
Norwegian (Nynorsk)	4	6,358	-







The PRINCIPLE Project then proceeded in Two Phases:

Data Provider		Country
National University of Ireland Galw	vay (NUIG)	Ireland
CIKLOPEA D.O.O		Croatia
celandic Ministry of Foreign Affair	S	Iceland
Standards Norway		Norway
Norwegian Ministry of Foreign Affa	airs	Norway
standards		RÁÐUNEYTIÐ ign Affairs Iceland
OÉ Gaillimh NUI Galway	NORWECIAN	

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Data Provider	Country
Rannóg an Aistriúcháin	Ireland
Foras na Gaeilge	Ireland
CIKLOPEA D.O.O	Croatia
Ministry of Foreign and European Affairs	Croatia
Icelandic Standards	Iceland
Icelandic Met Office	Iceland
Tithe an Oireachtais Houses of the Oireachtas	Veðurstofa Íslands
Íslenskir staðlar CIKLOPEA	Foras na Gaeilge



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Language Resources collected in PRINCIPLE - Croatian

Dataset	TUs Collected	Data used in MT
EN>HR Baseline	3,891,799	3,708,493
MVEP Data	115,667	100 640
Other Data Providers	22,703	- 100,649



REPUBLIC OF CROATIA Ministry of Foreign and European Affairs

Dataset	TUs Collected	Data used in MT
HR>EN Baseline	3,891,799	3,708,493
Ciklopea Data (eProcurement)	36,634	- 47,135
Other Data Providers	22,703	

Dataset	TUs Collected	Data used in MT
EN>HR Baseline	3,891,799	3,708,493
Ciklopea Data (eHealth)	76,108	72,455









Language Resources collected in PRINCIPLE - Irish

Dataset	TUs Collected	Data used in MT
EN>GA Baseline	901,421	588,663
Foras na Gaeilge	60,443	54,141
Rannóg an Aistriúcháin	387,480	353,485
Dept. Culture & Gaeltacht	64,694	58,057



Dataset	TUs Collected	Data used in MT
EN>GA Baseline	901,421	588,663
Rannóg an Aistriúcháin	387,480	353,485
Dept. of Justice	35,898	28,639
Dept. Culture & Gaeltacht	64,694	58,057





An Roinn Dlí agus Cirt Department of Justice

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Language Resources collected in PRINCIPLE - Icelandic

Dataset [EN<>IS]	TUs Collected	Data used in MT
Ministry of Foreign Affairs Data	1,097,352	821,243

Note that the Icelandic Ministry of Foreign Affairs stipulated only their data to be used, no baseline/other data.



UTANRÍKISRÁÐUNEYTIÐ Ministry for Foreign Affairs Iceland

Dataset	TUs Collected	Data used in MT	
IS>EN Baseline	801,283	702,139	
Icelandic Met Office Data	214,242	188,700	



Dataset	TUs Collected	Data used in MT
EN>IS Baseline	801,283	702,139
Standards Iceland Data	16,590	16,423







Language Resources collected in PRINCIPLE – Norwegian [Bokmål]

Dataset [EN>NO]	TUs Collected	Data used in MT
Norwegian Ministry Foreign Affairs	1,757,609	1,616,568



Note that the Norwegian Ministry of Foreign Affairs stipulated only their data to be used, no baseline/other data.

Dataset	TUs Collected	Data used in MT
EN>NO Baseline	1,964,961	1,140,351
Standards Norway Data	132,360	77,664











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An Overview of Automatic MT Evaluation in PRINCIPLE

For every MT model developed by Iconic, a sanity-check evaluation was conducted against freely available online MT Engines.



A test set of 2,000 segments is generally held out as a test from data provided by customers. In some cases with PRINCIPLE Early Adopters, where limited data was provided, a test set of 1,000 segments or 1,500 segments was used.



Test segments are run through multiple MT engines for comparison, with a range of metrics computed [SacreBLEU, TER, METEOR, chrF].

• Each data set (bar triplets) represents the evaluation on a held-out test set for that model, either a baseline model for the language (PRINCIPLE), or a model with Early Adopter data.



Comparing PRINCIPLE Engines to Online MT - Croatian



Comparing PRINCIPLE Engines to Online MT – Irish



Comparing PRINCIPLE Engines to Online MT – Icelandic



Comparing PRINCIPLE Engines to Online MT – Norwegian



Sample User Evaluations

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An Overview of User MT Evaluation in PRINCIPLE

Each PRINCIPLE 'Early Adopter' was invited to develop a test set to be used by DCU (Evaluation co-ordinator) to help evaluate MT both using automatic and manual means.



- A test set was requested of 500 segment pairs that
- Had not already been provided to train the MT systems.
- Were representative of the texts intended to be translated with the MT system.
- The reference translation in the target language should not be obtained via MT/Post-edit.
- Did not contain any confidential material.



Early Adopters were offered a range of human evaluation protocols from which they could choose, depending on their preference and available resources.

 Comparative ranking, adequacy & fluency, direct assessment, comprehension, post-editing, or MT error analysis





Comparison of MT Engines at Norwegian MFA [EN-NO]

A 500-segment Test Set was created by MFA Norway, separate from all training data.



An automatic evaluation was conducted independently by DCU of four MT engines.

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Comparison of MT Engines at Norwegian MFA [EN-NO]



A direct comparison of two engines was conducted by three evaluators at MFA Norway across the 500-segment test set (one evaluator completed only half of the test set).

For 70% of segments, Iconic's MFA engine was equal to or better than the comparator.

		ator 1 gments)		uator 2 egments)		uator 3 egments)		tal Igements)
Iconic Best	229	45.8%	260	52.0%	94	37.6%	583	46.6%
Online Best	138	27.6%	127	25.4%	68	27.2%	333	26.6%
Equally Good	118	23.6%	84	16.8%	86	34.4%	288	23.0%
Equally Poor	14	2.8%	29	5.8%	1	0.4%	44	3.5%
Not Assigned	1	0.2%	0	0.0%	1	0.4%	2	0.1%
Total	500	100%	500	100%	250	100%	1,250	99.8%





Comparison of MT Engines at Foras na Gaeilge [EN-GA]

A 496-segment Test Set was created by Foras na Gaeilge, separate from all training data.



An automatic evaluation was conducted independently by DCU of four MT engines.

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Evaluation of MT Output at Foras na Gaeilge [EN-GA]



Two FnaG translators undertook Adequacy and Fluency evaluation of Iconic MT output on the 496 test segments, using a 4-point Likert scale. The questions were

- How much of the information and meaning expressed in the source is conveyed accurately in the translation?
- How fluent is the translation?

Measurement of inter-translator agreement:

Cohen's Kappa	Adequacy	Fluency		
Non-weighted	0.009	0.011		
Weighted	0.031	0.026		

- Generally low agreement between translators
- Translator 2 more strict ratings 2-3, not 4

Translators' Rating of Adequacy and Fluency

	Adequacy	Fluency
Average	3.57	3.36
Mode	4	4





Comparison of MT Engines at Met Office Iceland [IS-EN]

A 500-segment Test Set was created by Met Office Iceland, separate from all training data.



An automatic evaluation was conducted independently by DCU of four MT engines.

MT Post-Editing at Met Office Iceland [EN-IS]



Two Met Office translators undertook a Post-Editing exercise, each translator post-editing the entire 500 segment test set.

	Total Time	Avg. per Sentence
Translator 1	00:48:04	00:05.7
Translator 2	00:39:51	00:04.7

TER scores were calculated to compare similarity of MT output and PE result to the original reference translation, and HTER measured how much post-editing was performed on the MT output.

TER (Reference)	Translator1	Translator2	hTER (PE)	Translator1	Translator2	
Iconic MT	22.7	22.7	Iconic MT	12.9	5.9	
PE	20.1	21.8	• Translator 2 performed fewer post-edits			





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Deployment of MT to PRINCIPLE Early Adopter Users



Each PRINCIPLE 'Early Adopter' was set up with access to the MT model trained on their data for day-to-day use during the course of the project.



PRINCIPLE Early Adopters all work within the same use-case: MT to be used in conjunction with translator review / post-editing in the translation workflow.



Almost 1 million words have been processed through PRINCIPLE MT engines during the course of the project.





Some Feedback from Translators at PRINCIPLE Early Adopters

"It did a good job at translating the text without much input from the translator"



"It is easier to move clauses around and correct terms and grammar rather than starting from scratch"

"Post-editing was by some distance faster than translating from scratch" "If the question to be answered in this testing procedure is whether the machine translation is helpful and saves time in this sort of translation, then the answer is "absolutely""

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