



# Supervised Machine Learning for Summarizing Legal Documents

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# Introduction

- Legal field: Thousands of decisions
  - ▶ 200 000/y in Canada
- Multilingual, multiple fields
- Access to specific decisions
  - ▶ Classification
- Access to relevant content
  - ▶ Summarization



# DecisionExpress™: Classification and Summarization

 <b>Information</b>  <b>Subject:</b> Skilled workers  <b>Conclusion:</b> <b>allowed</b>  <b>Judge:</b> <a href="#">Michel Beaudry</a> <b>Tribunal:</b> Federal Court  <b>Document Type:</b> Judgment   <a href="#">View Summary</a>	<a href="#">Mepham v. Canada ( Citizenship and Immigration) ( 2009 FC 1188 )</a>  <b>Headnote:</b>  <b>Topics:</b>  <b>Location(s):</b> <a href="#">UNITED KINGDOM</a>  <p>This is an application for judicial review of the Immigration and Refugee Act, S.C. 2001, c. 27 , by Lynn Mepham, an Immigration Officer at the Canadian High Commission in London. The Immigration Officer rejected her application for permanent resident class.</p> <p>Selection criteria, Educational credentials</p>
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<b>Introduction</b>	[1] This is an application for judicial review pursuant to subsection 72(1) of the Immigration Act), by Lynn Mepham (the Applicant), of a decision of an Immigration Officer (the Officer) at London, dated December 22, 2008, where the Officer rejected her application for permanent worker class.
<b>Context</b>	[4] The Applicant is a citizen of the United Kingdom and made an application for permanent skilled worker class.  [5] The Applicant submitted information pertaining to her education. She listed that she had

# Objective of our work

- Current system on *DecisionExpress*<sup>TM</sup>:
  - ▶ **Symbolic** method
  - ▶ Manual linguistic rules
- Our goal:
  - ▶ Does a **statistical** method can get similar performances?
  - ▶ Will the results be **constant** over legal fields and languages?

# DecisionExpress<sup>TM</sup>'s Architecture

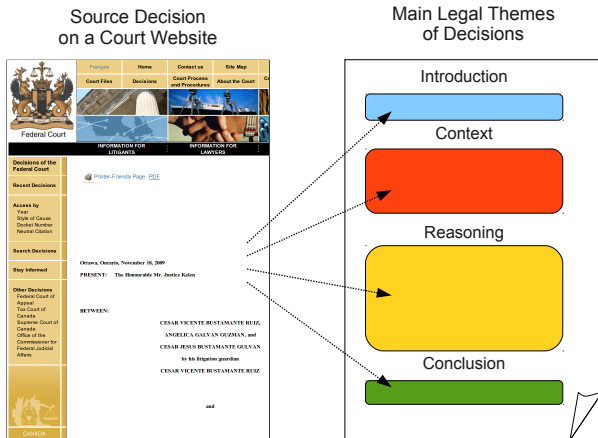
## HTML Decision on a Court's Website



Structure Analysis

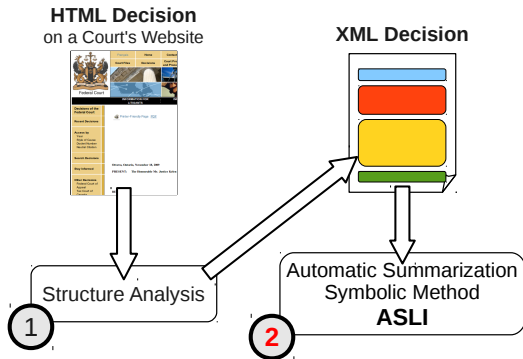
1

# Automatic Structure Analysis

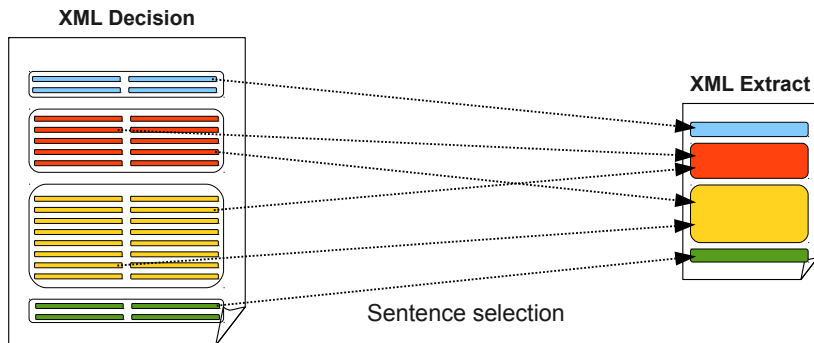


- Structure elements: sections (legal themes), paragraphs, sentences

# DecisionExpress<sup>TM</sup>'s Architecture



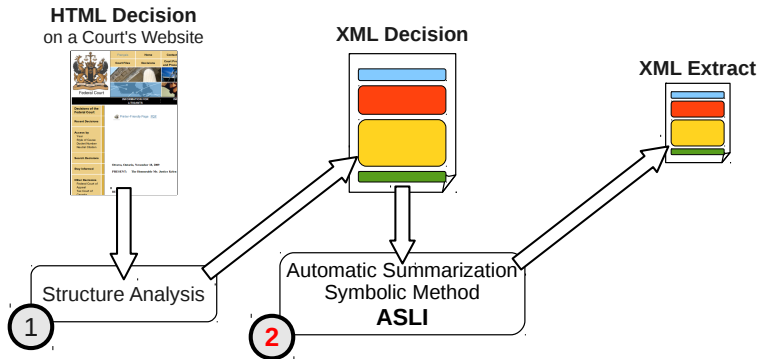
# Extracting Sentences



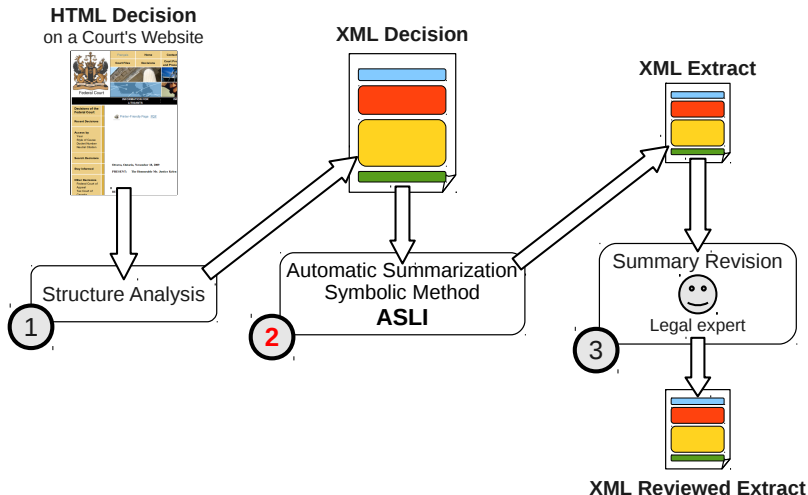
- Select whole sentences
- Sentences may fall into another section



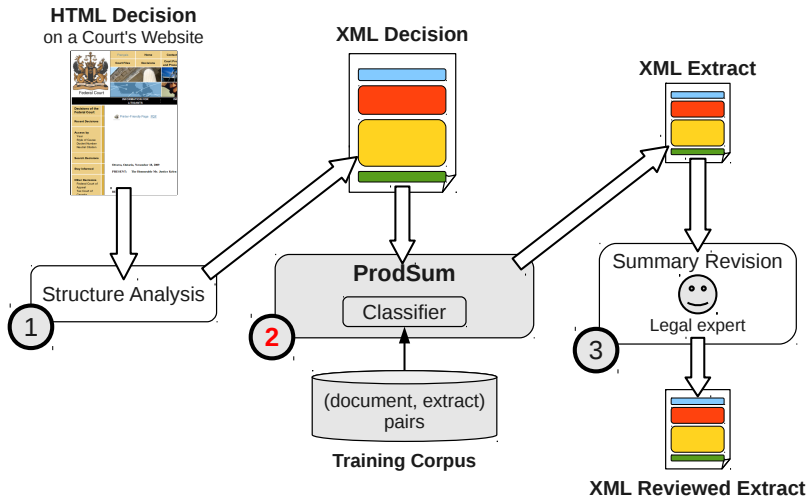
# DecisionExpress™'s Architecture



# DecisionExpress<sup>TM</sup>'s Architecture



# DecisionExpress<sup>TM</sup>'s Architecture: Our Objective

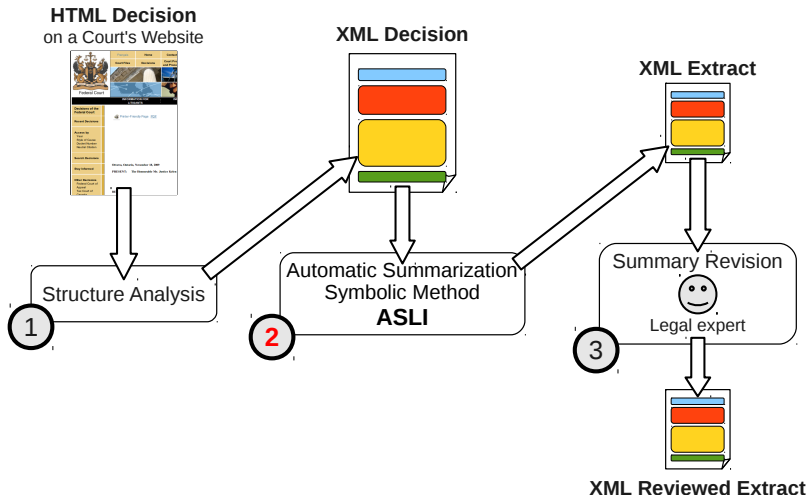


- 1 year of activity on *DecisionExpress*<sup>TM</sup>
  - ▶ ~ 4000 decision and extract pairs

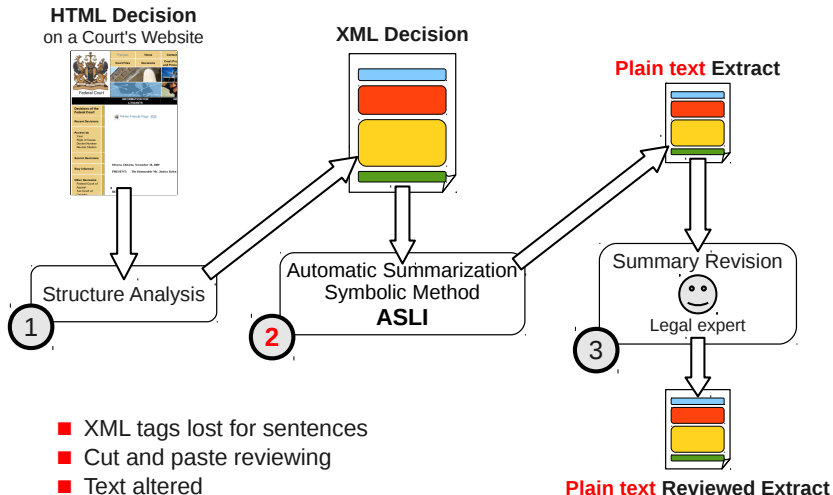
Field	IMM	TAX	IP
English	<b>1 765</b>	447	176
French	1 155	164	8

- Issue: Extract's sections were reviewed in plain text

# DecisionExpress™'s Architecture

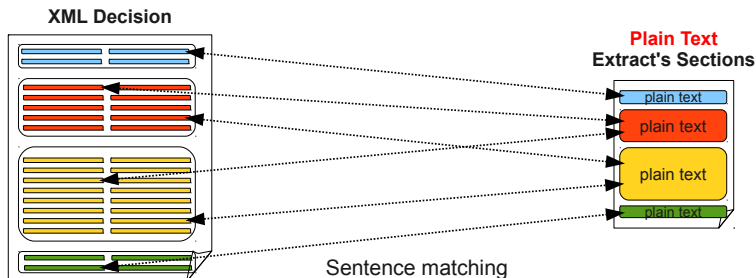


# DecisionExpress™'s **Previous** Architecture



# Corpus' Extracts: Plain Text to XML

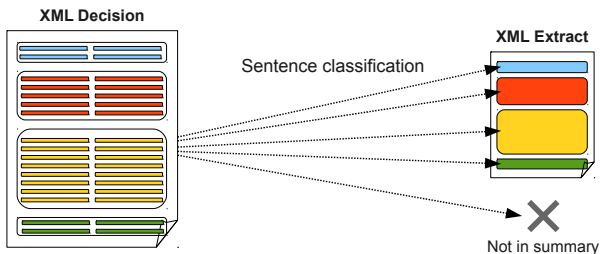
- **Goal:** Find extract sentences that match the source ones



- **Issues:** Sentence alterations, merging, splitting, reordering...
- **Algorithm:** String alignment allowing distance editing
- **Result:** 94% of matched sentences

# Machine Learning

- Classifier: Naive Bayes
- Instances: Sentences
- Training set: 2/3rd
- 5 classes: Selected for a section or “not in summary”





# Classification Features

## Surface

Position, length, count of sentences, paragraphs, sections.

## Emphasis

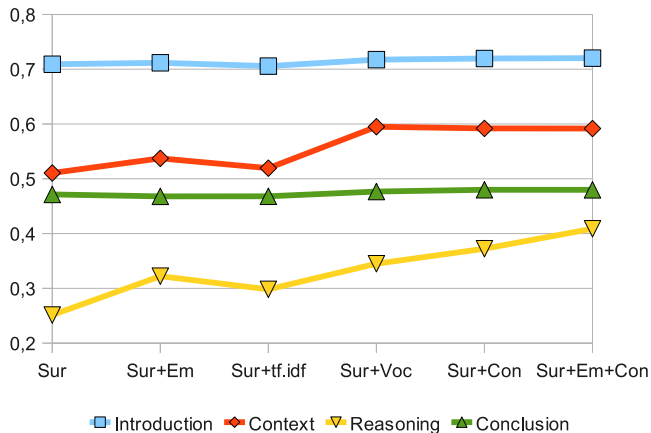
HTML styles: bold, underline, italic, indent.

## Content

- $tf \cdot idf$  normalized sum
- “Extract’s specific vocabulary” score
  - ▶ High score: “apparently”, “dismissed”, “daughter”, “kill”...
  - ▶ Low score: “paragraphs”, “relies”, “procedure”...

# Classification Results for English Immigration Decisions

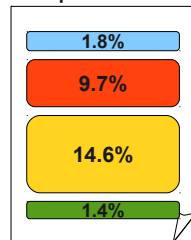
F1-Measure



# Comparison with a baseline and ASLI

- **PRODSUM**: Surface + Emphasis + Content
- **ASLI**: Current symbolic system
- **Baseline**: First sentences of each section based on the average compression ratio

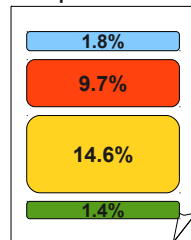
Average Extract's  
Compression Ratio



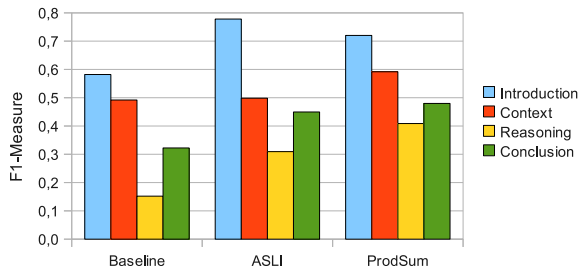
# Comparison with a baseline and ASLI

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Average Extract's  
Compression Ratio



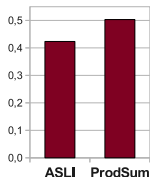
System classification scores



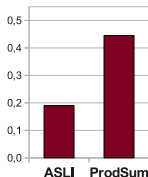
# Comparison over legal fields and languages

## English

### Immigration

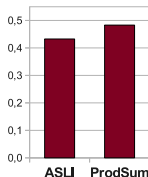


### Tax

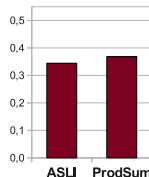


## French

### Immigration



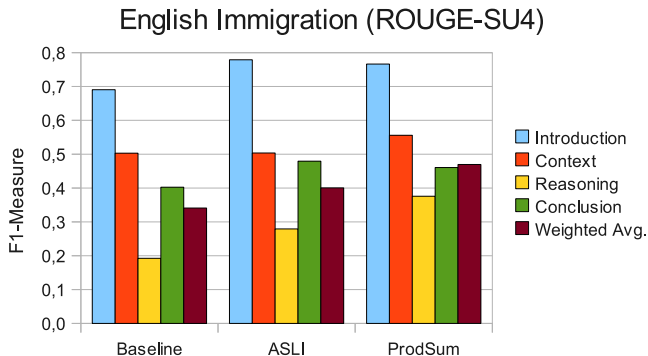
### Tax



Weighted average F1-Measure scores of all sections

- ASLI's rules do not adapt well to the tax field
- PRODSUM is more reliable over fields

# Content evaluation: ROUGE scores



- No control over extract's sizes  $\Rightarrow$  F1-Measure scores
- PRODSUM works best for longer sections
- Introduction and Conclusion are better handled with ASLI's rules

- Corpus of 4000 decisions suitable to ML
- Emphasis and “extract’s vocabulary” features improve results
- Basic features generally beat the symbolic method
- PRODSUM adapts better to new legal fields

## Perspectives

Features based on events, entities and factual information.

Thank you for your attention.

Questions, comments?