

AUTOMATED INTERPRETATION

THE OTHER, BETTER, WISER AI

- Machine Learning models often claim to predict rules by analysing data and answers.
- This is empirical analysis to an extreme, but risks omitting extraneous data created from unpredictable factors like human behaviour.

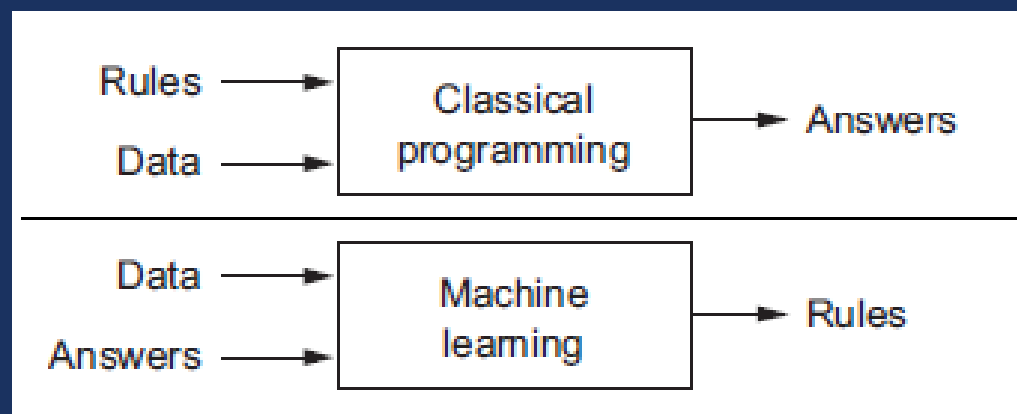


Figure 1 – A comparative model of procedural programming and machine learning.

- By observing seminars as case studies, and evaluating them as qualitative data, it was possible to gain insight into socio-industrial trends and methods through the perspective of workers within that context.

- These seminars can offer a large set of rich qualitative data, by looking at data taken from individual seminars and looking at the meta data of the collective set.

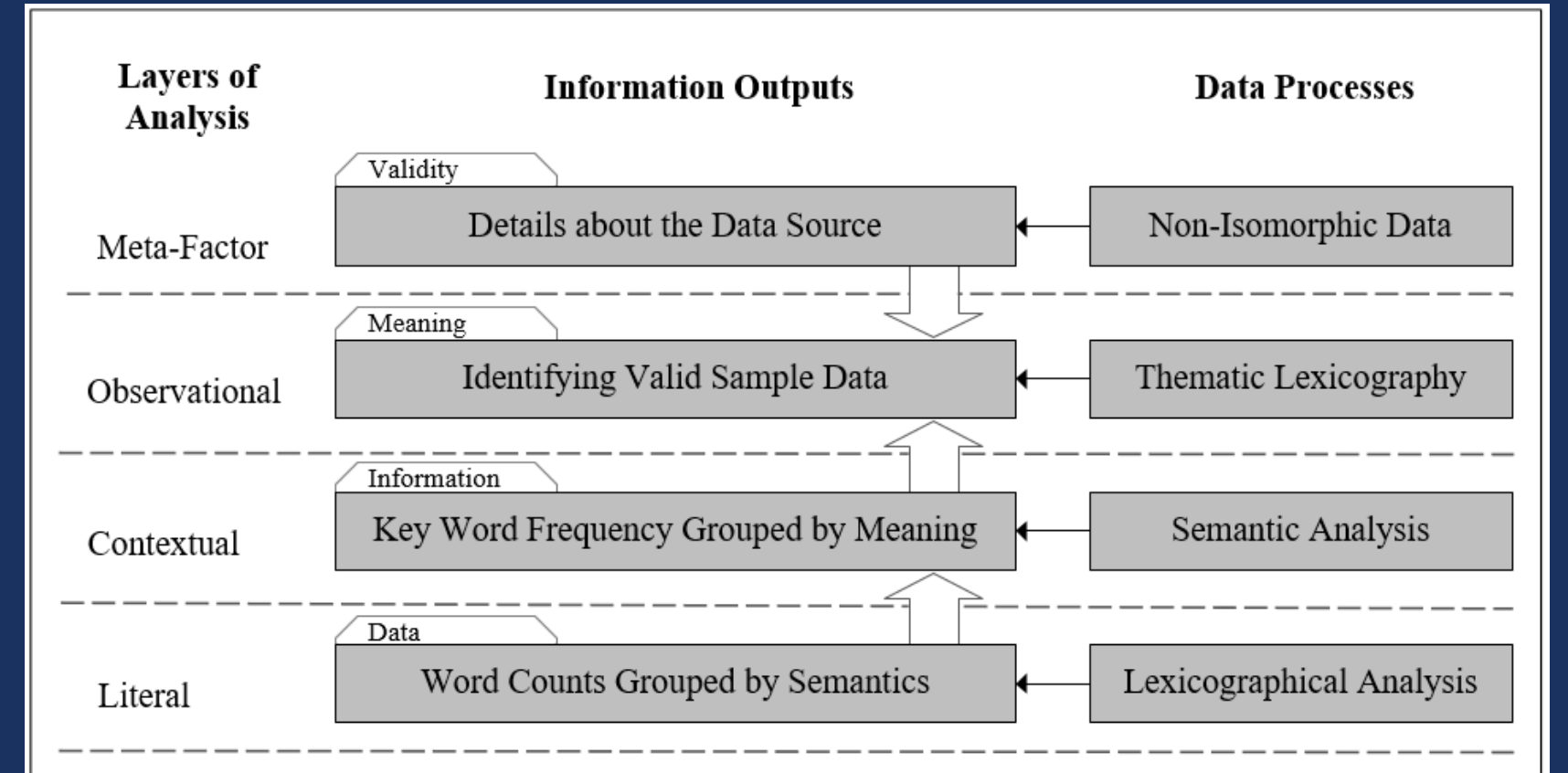
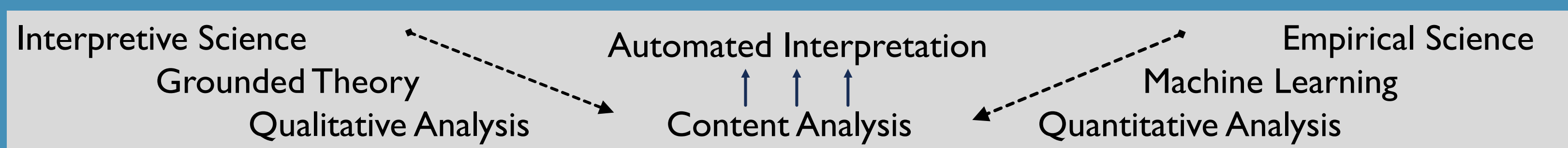


Figure 2 – A systematic, multi-modal model of interpretive content analysis for application of deep learning for complex qualitative data sets.

Problem Analysis and Formative Challenges



A. Limitations of Automated Qualitative Analysis

- Expensive to initially gather data
- Processing and analysis is highly subjective
- Hard to replicate, hard to control

Validation

B. Limitations of Procedural Quantitative Analysis

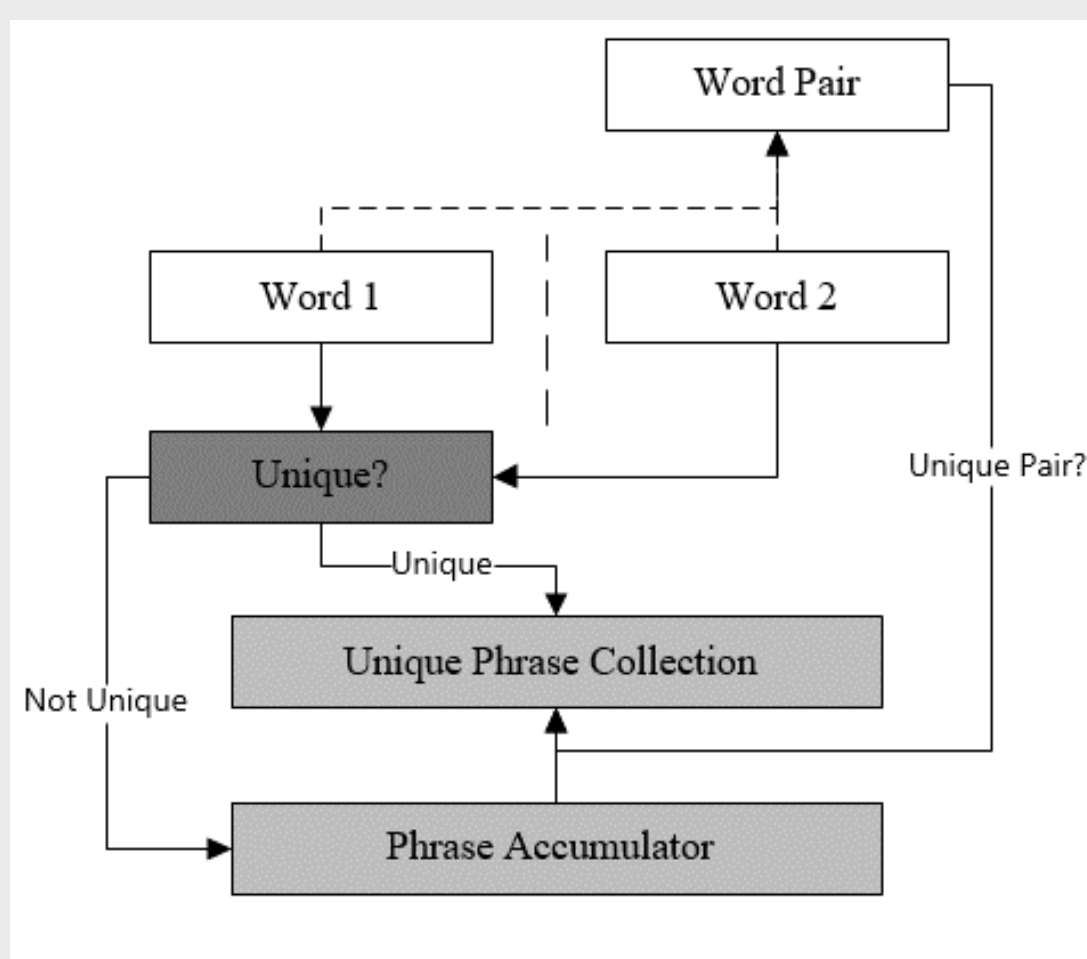
- Abstract and prone to omit non-isomorphic data
- Predisposed to unobserved biases and
- Can lacks semantic and contextual understanding

Verification

New Methods of Content Analysis and Experimental Findings

1. Literal Data Modelling

Lexicographical analysis and linear regression are applied to high-level data from the sample (such as seminar titles) and used to identify localised patterns within the data set.

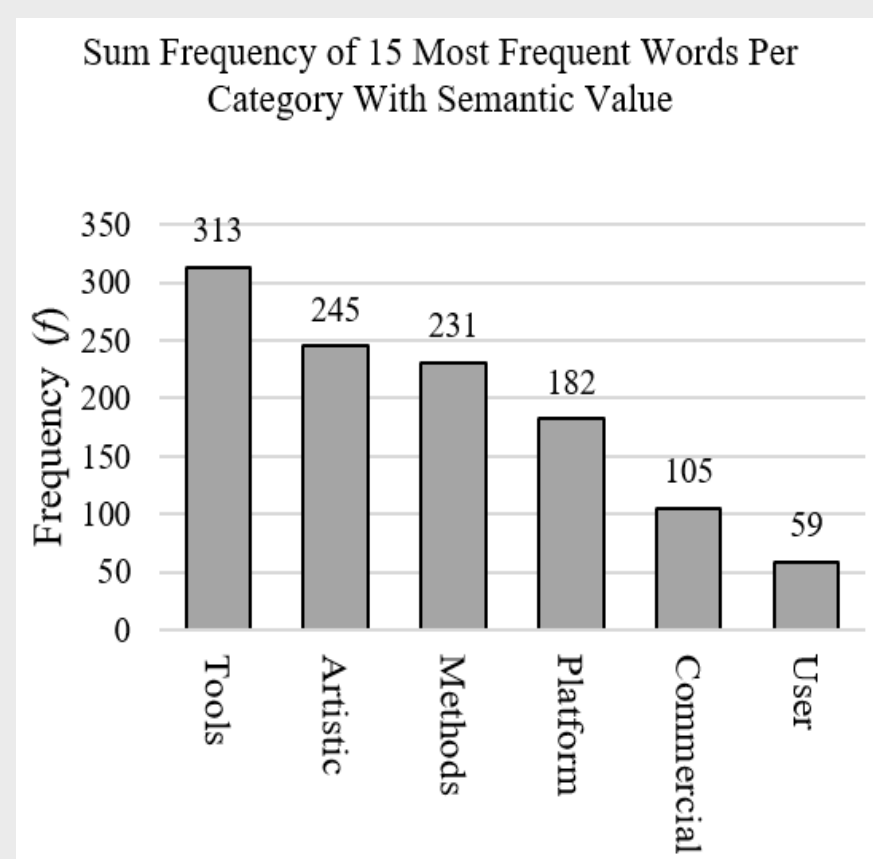


Tip: Lexicography models are not best developed in isolation. A representative sample of professional experts or should help to train dictionary models for a given specialised corpus.

Figure 3 – A word-pairing algorithm combined with natural language processing is used to model key trends and factors at a high-level from the data set.

3. Contextual Language Modelling

Deep learning is applied to the full content of each qualitative data set to model key topics to form a corpus-based language model.



Tip: Dictation programs were found to be up to 37% more accurate than auto-captioning libraries.

'Big Data' on the cloud has allowed language process of timbre, accents, and intonation from speaker(s) to improve analysis and recognition.

Figure 5 – A corpus-based semantic model is trained to recognise aggregated trends.

2. Observational Data Modelling

Word pairs are semantically grouped and modelled to find aggregate trends across the samples using recursive analysis.

2017				2018				2019			
Ch	Rank	Δ	Concat	Ch	Rank	Δ	Concat	Ch	Rank	Δ	Concat
3	19	▲	Animation	1	23	▲	Reflection	1	14	▲	Your
3	19	▲	Indie	2	15	▲	Player	1	14	▲	Reflection
9	8	▲	Generation	3	14	▲	Your	2	9	▲	Art
9	8	▲	Procedural	4	13	▲	Story	3	7	▲	Creating
10	7	▲	Day	5	12	▲	Making	3	7	▲	Building
10	7	▲	No	5	12	▲	Art	3	7	▲	Bootcamp
10	7	▲	LUX	6	11	▲	Creating	4	6	▲	Making
2	21	▲	Your	6	11	▲	Animation	4	6	▲	Play
8	10	▲	Character	7	10	▲	Level	4	6	▲	World
11	6	▲	Direction	7	10	▲	Day	4	6	▲	Player
11	6	▲	Clur	7	10	▲	AI	4	6	▲	AI
11	6	▲	Programmer	7	10	▲	Building	5	5	▲	Your Game
11	6	▲	Social	7	10	▲	Development	5	5	▲	Level

Figure 4 – A recursive model is used to identify initial trends over aggregate data.

Tip: Compile a lexicography filtering out stop words and then compile one which retains them, analysing both. Stop words lack semantic meaning when analysed in isolation, but add details to word pairs. E.g. 'Your game' is prescriptive, but 'our game' is reflective.

4. Meta-factor Data Source Modelling

Observations data-mined from the data source offers rich details to facilitate multiple-linear regression to identify trends, while maintaining power of the experiment as the sample size is further regressed and shrinks.

Demographic				Tooling			
Game Dev	Research	Count	Phrase	Count	Phrase	Count	Phrase
30 Bug	12 Proc-Gen	44	Rig	44	Rig		
25 Player	13 Proc-Gen	44	Rig	44	Rig		
18 Designer	10 Visual	44	Rig	44	Rig		
18 Build	10 Visual	44	Rig	44	Rig		
17 Simple	10 Visual	44	Rig	44	Rig		
14 Editor	10 Visual	44	Rig	44	Rig		
14 Test	10 Visual	44	Rig	44	Rig		
14 QA	10 Visual	44	Rig	44	Rig		
13 Artist	10 Visual	44	Rig	44	Rig		
12 Data	10 Visual	44	Rig	44	Rig		
12 Trigger	10 Visual	44	Rig	44	Rig		
11 Fix	10 Visual	44	Rig	44	Rig		
11 Designer	10 Visual	44	Rig	44	Rig		

Tip: There is always non-isomorphic data which can't be anticipated or which will be tainted by observed or unobserved biases.

Figure 6 – The corpus-based semantic model is cross-referenced against meta-data from the source to find deeper trends from the aggregate data.