

Spare us the surprise

The interplay of paradigmatic predictability and frequency

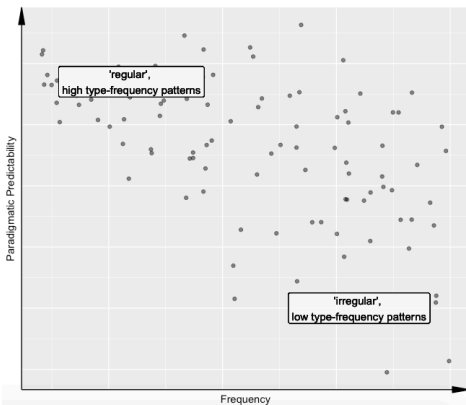
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ISMo 2021

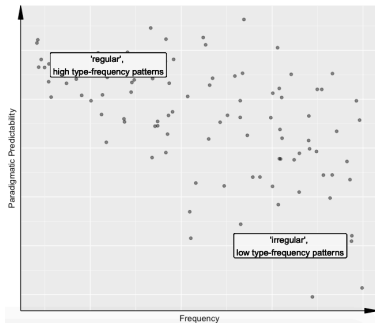
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Introduction

- A known inverse link between **frequency and paradigmatic predictability** of a word form (Wu, Cotterell & O'Donnell, 2019; Marcus et al. 1992; Bybee, 1985):
 - Paradigmatically unpredictable word forms (suppletives/irregulars) tend to be frequent
 - Infrequent lexemes tend to have predictable word forms

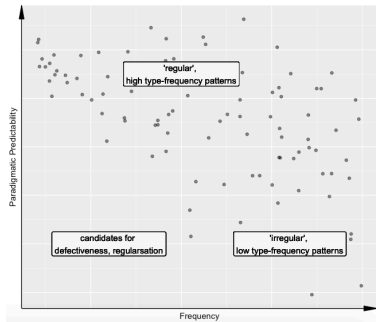


Uncertainty, frequency and memory



- The more **high frequency** a word form, the more it can **afford to be paradigmatically unpredictable**.
 - The unpredictable word form can be well anchored in memory thanks to its frequency

Uncertainty, frequency and memory



- If a **paradigmatically unpredictable word is infrequent/in an infrequent context...**
 - Regularisation (Eng. *helped* ← *holp*) (Lieberman et al. 2007)
 - Avoidance (*forego* → *foregoed?*/*forewent?*) (Albright, 2003; Sims, 2015)
 - If a whole context is infrequent and a locus of low predictability, it may drop out of use (It. *passato remoto*)

- A **negative correlation** between frequency and predictability
- For an **unpredictable** form to survive, it must be **frequently attested**
- Words **can't afford to be both syntagmatically and paradigmatically unpredictable** (Filipović Đurđević & Milin, 2019)
 - Frequent words are an expected way to continue a sentence (= syntagmatically more predictable), so they can tolerate paradigmatic uncertainty.

The effect of paradigmatic predictability on speaker production

- When producing a sentence, we incrementally have to find words that
 1. are **inflectionally appropriate** (e.g. have the correct agreement, are the correct part of speech: *she eats/*eat dinner*)
 2. are **an appropriate lexical choice** (contribute the intended lexical semantics: *she eats/*coagulates dinner*)

The effect of paradigmatic predictability on speaker production

- At each word boundary, the inflectional requirements are often clear
 - *"You should beware of the dog!" - "Indeed, yesterday, I _____ (it)!"* needs a past tense form.
- Several appropriate lexemes (HEED, BEWARE OF, WATCH OUT FOR). Some parameters for the choice:
 - Overall strength of the lexeme's mental representation (a function of recency, frequency in input, salience given context...)
 - **The ease of accessibility of the necessary form of the lexeme** (a function of predictability)

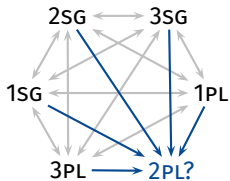
- How does paradigmatic predictability impact token frequency?
- **The hypothesis:**
 - at parity of lexeme frequency, less paradigmatically predictable words will be used less frequently.
 - The more frequent a lexeme, the less predictability will matter for frequency of use (frequent words need to be retrieved from memory rather than actively predicted)

Aspects of form predictability

- **Several aspects of form predictability** may be relevant to token frequency, e.g.:
 1. **Local entropy**: the uncertainty surrounding how to fill a given cell
 - PRS *fling* → PST $\left\{ \begin{array}{l} \textit{flung?} \\ \textit{flang?} \\ \textit{flinged?} \end{array} \right.$
 2. **Surprisal**: the predictability of the particular form actually filling the cell
 - PRS *fling* → PST *flung*
- Following a corpus study, we conclude that the measure relevant for written production is surprisal **How to operationalise?**

Form predictability as average surprisal i

- Need to measure:
 - given **knowledge of the rest of the paradigm**...
 - how **confident** should a speaker be that they are **producing the right form in the necessary cell**?
- This is clearly a variant of the **Paradigm Cell Filling Problem** (Ackerman, Blevins & Malouf, 2009; Ackerman & Malouf, 2013).



- We rely on a purely **word-based** approach to the PCFP of Bonami & Beniamine (2016)

Operationalising surprisal

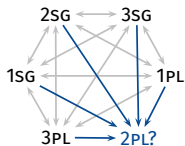
- Beniamine's (2018) Qumin package was used for all computations.
- Intuitively: **conditional probability of output form given the phonological shape of the input form**
- Surprisal is computed over **pairs of cells** ($C1 \rightarrow C2$). For a given form pair...
 - Find all patterns compatible with the input form
 - $x = \frac{\text{type freq. of instantiated pattern}}{\text{type freq. of all applicable patterns}}$
 - Turn it into bits: $-\log_2(x)$

1PL IND PRES *sortons* → PST PART $\left\{ \begin{array}{l} \textit{sorté?} \\ \textit{sortu?} \\ \textit{sorti?} \end{array} \right.$

PATTERN	PATTERN TYPE FREQUENCY	SURPRISAL
<i>Xons ~Xé</i>	most lexemes	0.06
<i>Xons ~Xu</i>	~ 15 lexemes	4.7
<i>Xons ~Xi</i>	~ 5 lexemes	7.2

Average surprisal

- **Average over predictor cells** c to get an overall estimation of how surprising c' is given the rest of the paradigm.



- Ideally, this should be **weighted by cell frequency**.
 - But we do not have quality estimations of cell frequency, because of pervasive syncretism.
 - For lack of a better solution we use unweighted frequency.

- We set out to confirm that **paradigmatic surprisal** has a **negative effect on token frequency** throughout the lexicon.
- And that the effect is **reversed for high-frequency lexemes**.
- Case study: French verbal cells

- For the items within each cell, we constructed a model of the shape
 - $\text{token frequency} \sim \text{surprisal} + \text{lexeme frequency} + \text{surprisal}:\text{lexeme frequency}$
- The value of surprisal we employ is the **average surprisal** of the given form based on each of the other forms in the paradigm.
- Lemma frequency is included as a control variable (= familiarity)
- The interaction: test the intuition that **for high values of lemma frequency, surprisal matters less** (words with a strong representation in memory don't need to be predicted)
- Separate bayesian poisson regressions with weakly-informative priors were fitted to the data in each cell.

- Resources used:
 - Frequency counts: FrCoW (Schäfer & Bildhauer, 2016) for token and lemma counts.
 - Paradigms & excluding homographs: GLàFF (Hatout, Sajous & Calderone, 2014)
 - Surprisal: values computed using Qumin (Beniamine, 2018) on the Flexique verb dataset (Bonami, Caron & Plancq, 2014)

Data selection

- Which cells in the paradigm of French verbs can we work with?
- Working with our dataset, we exclude...

Finite forms						
	1SG	2SG	3SG	1PL	2PL	3PL
IND.PRS	2	3	183	2	5	14
IND.IPFV	0	0	5083	10	10	5076
IND.PST	4484	4448	4694	5116	5116	5101
FUT	5211	5207	5213	5190	5212	5221
SBJV.PRS	0	250	2	8	7	13
SBJV.IPFV	4701	4725	5119	4726	4738	4740
COND	0	0	5220	5212	5212	5215
IMP	—	0	—	2	2	—

Nonfinite forms					
INF	PRS.PTCP	PST.PTCP			
		M.SG	F.SG	M.PL	F.PL
5006	4311	3935	3055	2903	3199

Number of verbs from Flexique with no homograph documented in the GLÀFF, by paradigm cell

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 - cells with high numbers of homographs according to the GLÀFF;
 - cells out of current usage (i.e. most attestations are likely to be archaic);
 - past participle cells, for which tagging is inherently unreliable.

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Number of verbs from Flexique with no homograph documented in the GLÀFF, by paradigm cell

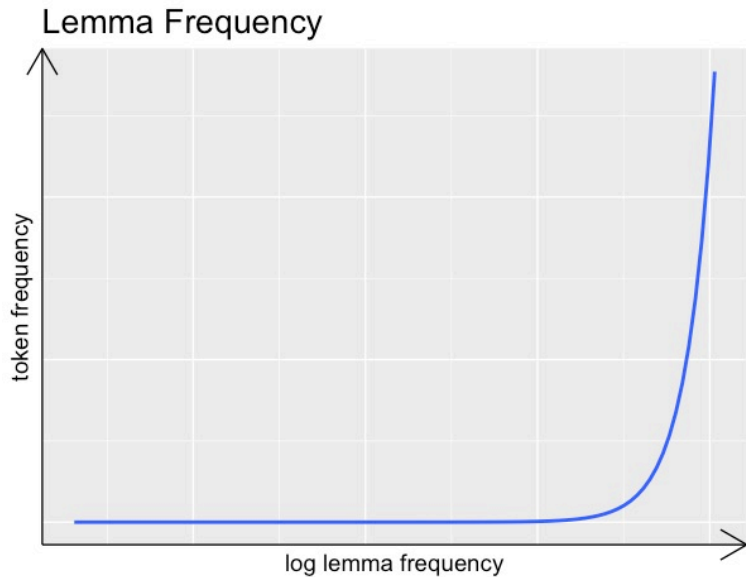
Properties of the selected cells

- The selected cells correspond to 3 areas of high interpredictability.

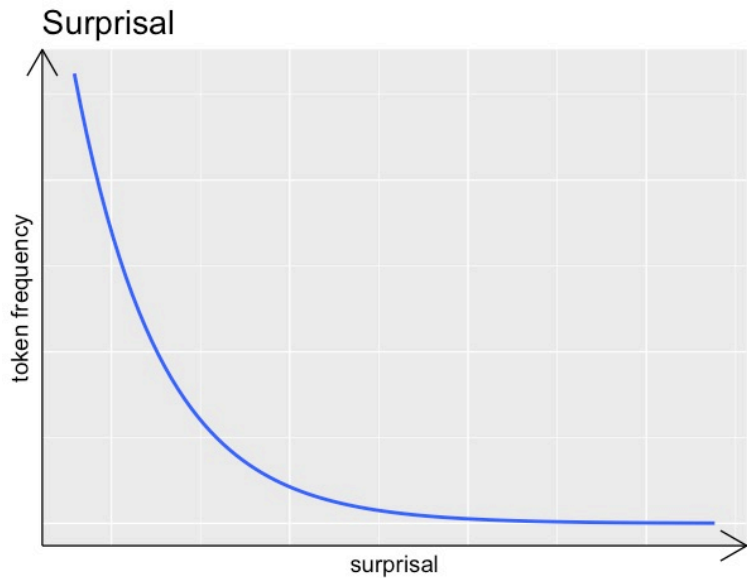
FUT.1SG	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
FUT.2SG	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
FUT.3SG	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
FUT.1PL	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
FUT.2PL	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
FUT.3PL	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
COND.3SG	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
COND.1PL	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
COND.2PL	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
COND.3PL	0	0	0	0	0	0	0	0	0	0	0.24	0.24	0.24	0.23
IPFV.3SG	0.35	0.34	0.34	0.34	0.35	0.34	0.35	0.33	0.33	0.35	0	0	0.0004	0.34
IPFV.3PL	0.35	0.34	0.34	0.34	0.35	0.34	0.35	0.33	0.33	0.35	0	0	0.0004	0.33
PRS.PTCP	0.34	0.34	0.34	0.34	0.34	0.34	0.34	0.33	0.33	0.34	0	0	0	0.32
INF	0.059	0.059	0.059	0.059	0.059	0.059	0.059	0.059	0.059	0.059	0.076	0.077	0.074	0
	FUT.1SG	FUT.2SG	FUT.3SG	FUT.1PL	FUT.2PL	FUT.3PL	COND.3SG	COND.1PL	COND.2PL	COND.3PL	IPFV.3SG	IPFV.3PL	PRS.PTCP	INF

Implicative entropy (Bonami & Beniamine, 2016) between selected cells

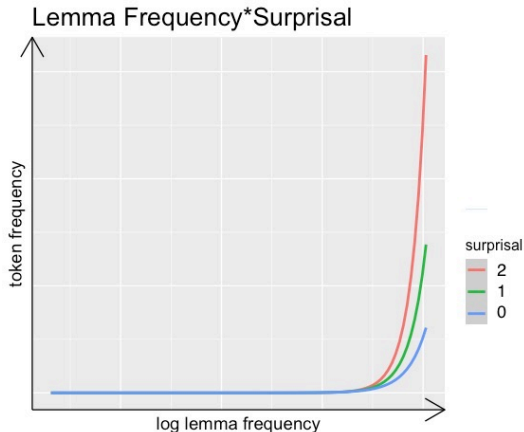
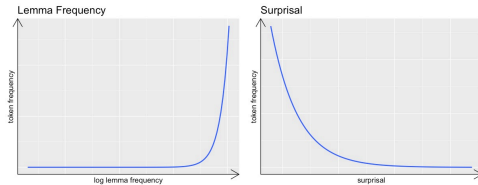
Predictions



Predictions



Predictions



Results

- **Lemma frequency** has a uniform positive effect on token frequency in all cells.
- **Surprisal** had a negative effect in 12/14 cells, an effect indistinguishable from 0 in 1/14, and an unexpected positive effect in 1/14.
- **The interaction between surprisal and lemma frequency** had a positive coefficient in 11/14 cells and an effect indistinguishable from 0 in 1/14. 2/14 have unexpected negative coefficients.
- **Overall, 11/14 cells behaved exactly as predicted, two behaved counter to expectations and one showed non-significant impact for surprisal and surprisal:lemma**

Model Output - Coefficients

Cell	Lemma freq.	Surprisal	Interaction
FUT.1SG	0.9935	-0.3783	0.0675
FUT.2SG	1.0771	-0.2306	0.0447
FUT.3SG	1.1764	-0.0261	0.0073
FUT.1PL	0.9693	-0.1932	0.0415
FUT.2PL	1.1072	-0.3368	0.0647
FUT.3PL	1.1466	-0.0040	0.0088
COND.3SG	1.2509	-1.0392	0.1835
COND.1PL	1.2544	-1.7739	0.2876
COND.2PL	1.2583	-2.7622	0.4486
COND.3PL	1.2312	-1.3889	0.2404
IPFV.3SG	1.1707	-0.0441	-0.0010
IPFV.3PL	0.9352	-0.5588	0.0959
PRS.PTCP	0.5916	0.0545	0.0053
INF	0.9438	0.0620	-0.0089

Unexpected coefficient sign

95% Credible interval overlaps with zero

- **Cells that didn't conform to predictions:** infinitive, imperfect 3sg, present participle.
- These are by far the three most frequent cells in the dataset.
- Hypothesis: the effect of surprisal is therefore nullified at the level of the whole cell (same mechanism for frequent lexemes)
 - while the coefficients for surprisal and the interaction have unexpected monotonicity, their value is much smaller compared to other cells, and very close to 0 (for pres. part. it is indistinguishable from 0)

- Overall, **token frequency is negatively impacted by paradigmatic form predictability**.
 - The pattern is reversed for items of high lemma frequency.
 - High frequency lexemes are **more familiar to speakers, so the predictability of their word forms matters less** for access/usage
- The method performs well on 11/14 cells, and the exceptions exist for principled reasons.
 - Showcases the **importance of paradigmatic information** in predicting frequency.
 - Frequent contexts and lexemes diminish the importance of paradigmatic predictability.

What next?

- Obtaining a **good estimate of cell frequency** (existing resources yield poor estimates, especially for the person dimension)
 - It would allow a weighed average of surprisal to be used
 - It would help interpret outlying results.
- Currently exploring the **Italian verbal system with the same method** (less homography)
 - Results going in **roughly the same direction**, some kinks to iron out
- Testing the general effect of surprisal **psycholinguistically**.
 - Speakers appear sensitive to paradigmatic surprisal between individual nonwords.

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