

# Case and inflection classes: Testing the Form-Frequency-Hypothesis

Matías Guzmán Naranjo & Laura Becker

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# The Form-Frequency-Hypothesis

- (1) Whenever a category A is expressed by more phonological material than category B, then category A is less frequent than category B.
- Whenever category A is more frequent than category B, then category A will be expressed by a shorter or equally long form than category B.

The FFH as a causal implication between frequency and form requires that:

- **ALL** frequency asymmetries result in coding asymmetries in the expected direction, or at least in no asymmetry.
- **ALL** coding asymmetries also have a frequency asymmetry in the expected direction, or at least no asymmetry.

This in turn implies that

- the form-frequency effect must be observable in individual languages.
- Otherwise, observing form-frequency effects cross-linguistically could be due to a correlation with no direct causal relation (i.e. both phenomena follow from a different explanation).

# Why case and inflection classes?

What we want to test:

- Do language-specific coding asymmetries correlate with the expected frequency asymmetry?
- Can this correlation be observed across languages?

NB: We want to avoid the confound of semantic complexity and ‘markedness’.

Therefore, **case** and, more specifically, **inflection classes** are a suitable testing ground:

- There is no straightforward reason why, a given case should be semantically more complex than another one.
- Inflection classes have markers of different lengths within a single language.
- Inflection classes have different frequencies.
- Cases and their exponents are easy to search for in an annotated corpus.

# Inflection classes

- ‘An inflectional class is a set of lexemes whose members each select the same set of inflectional realizations.’ (Aronoff, 1994, p. 64)
- Inflection classes are morphemic, i.e. they are only relevant to morphology, whereas phonology, syntax, or semantics are not sensitive to them.

An example: Gothic *-a* declension

Day				Word				
	Singular	Plural		Singular	Plural			
Nom	dags	–s	dagōs	–ōs	waúrd	–	waúrda	–a
Acc	dag	–	dagans	–ans	waúrd	–	waúrda	–a
Gen	dagis	–is	dagē	–ē	waúrdis	–is	waúrdē	–ē
Dat	daga	–a	dagam	–am	waúrda	–a	waúrdam	–am

# Gothic inflection classes, *-ja* declension

		army				herdsman			
		Singular		Plural		Singular		Plural	
Nom	harjis	-jis	harjōs	-jōs	haírdeis	-eis	haírdjōs	-jōs	
Acc	hari	-i	harjans	-jans	haírdi	-i	haírdjans	-jans	
Gen	harjis	-jis	harjē	-jē	haírdeis	-eis	haírdjē	-jē	
Dat	harja	-ja	harjam	-jam	haírdja	-ja	haírdjam	-jam	

race				
		Singular		Plural
Nom	kuni	-i	kunja	-ja
Acc	kuni	-i	kunja	-ja
Gen	kunjis	-jis	kunjē	-jē
Dat	kunja	-ja	kunjam	-jam

# Gothic inflection classes, -ō declension

gift				
	Singular		Plural	
Nom	giba	-a	gibōs	-ōs
Acc	giba	-a	gibōs	-ōs
Gen	gibōs	-ōs	gibō	-ō
Dat	gibái	-ái	gibōm	-ōm

# Gothic inflection classes, *-jō* declension

band			
	Singular	Plural	
Nom	bandi	-i	bandjōs -jōs
Acc	bandja	-ja	bandjōs -jōs
Gen	bandjōs	-jōs	bandjō -jō
Dat	bandjái	-jái	bandjōm -jōm

# Gothic inflection classes, *-i* declension

stranger				wife				
	Singular	Plural		Singular	Plural			
Nom	gasts	-s	gasteis	-eis	qēns	-s	qēneis	-eis
Acc	gast	-	gastins	-ins	qēn	-	qēnins	-ins
Gen	gastis	-is	gastē	-ē	qēnáis	-áis	qēnē	-ē
Dat	gasta	-a	gastim	-im	qēnái	-ái	qēnim	-im

# Gothic inflection classes, *-u* declension

	son				property	
	Singular		Plural		Singular	
Nom	sunus	-us	sunjus	-jus	faíhu	-u
Acc	sunu	-u	sununs	-uns	faíhu	-u
Gen	sunáus	-áus	suniwē	-iwē	faíháus	-áus
Dat	sunáu	-áu	sunum	-um	faíháu	-áu

# Gothic inflection classes, *-an* declension

		man				heart			
	Singular	Plural		Singular		Plural			
Nom	guma	-a	gumans	-ans	haírtō	-ō	haírtōna	-ōna	
Acc	guman	-an	gumans	-ans	haírtō	-ō	haírtōna	-ōna	
Gen	gumins	-ins	gumanē	-anē	haírtins	-ins	haírtanē	-anē	
Dat	gumin	-in	gumam	-am	haírtin	-in	haírtam	-am	

# Gothic inflection classes, *-ōn* declension

tongue				
	Singular		Plural	
Nom	tuggō	–ō	tuggōns	–ōns
Acc	tuggōn	–ōn	tuggōns	–ōns
Gen	tuggōns	–ōns	tuggōnō	–ōnō
Dat	tuggōn	–ōn	tuggōm	–ōm

# Gothic inflection classes, *-ein* declension

wisdom				
	Singular		Plural	
Nom	frōdei	-ei	frōdeins	-eins
Acc	frōdein	-ein	frōdeins	-eins
Gen	frōdeins	-eins	frōdeinō	-einō
Dat	frōdein	-ein	frōdeim	-eim

# Gothic inflection classes, -r declension

sister				
	Singular		Plural	
Nom	swistar	-ar	swistrjus	-rjus
Acc	swistar	-ar	swistruns	-runs
Gen	swistrs	-rs	swistrē	-rē
Dat	swistr	-r	swistrum	-rum

# Gothic inflection classes, *-nd* declension

friend				
	Singular		Plural	
Nom	frijōnds	-s	frijōnds	-s
Acc	frijōnd	-	frijōnds	-s
Gen	frijōndis	-is	frijōndē	-ē
Dat	frijōnd	-	frijōndam	-am

## Gothic inflection classes

Looking at, e.g., the nominative and accusative singular forms of nouns from two inflection classes in Gothic, we see inverted lengths of the case markers:

- (2) -ja declension (Gothic)

  - a. *har-jis* 'army-NOM'
  - b. *har-i* 'army-ACC'

(3) -jō declension (Gothic)

  - a. *band-i* 'band-NOM'
  - b. *band-ja* 'band-ACC'

→ Can the length differences be explained by the frequencies of the forms?

## Testing the FFH: The present approach

# Corpora

We are using the Universal Dependencies Corpora (Silveira et al., 2014) for the following languages:

- Gothic (45k words)
- Latin (440k words)
- Ancient Greek (380k words)
- Modern Greek (51k words)
- Russian (1m words)
- Czech (1.3m words)
- Polish (72k words)
- Turkish (only used for comparison, 46k words)
- Latvian (only used for comparison, 44k words)
- Lithuanian (only used for comparison, 40k words)

We chose these languages based on their availability in the corpus, their case paradigms and inflection classes.

# Inter-corpora comparison

## A comparability problem

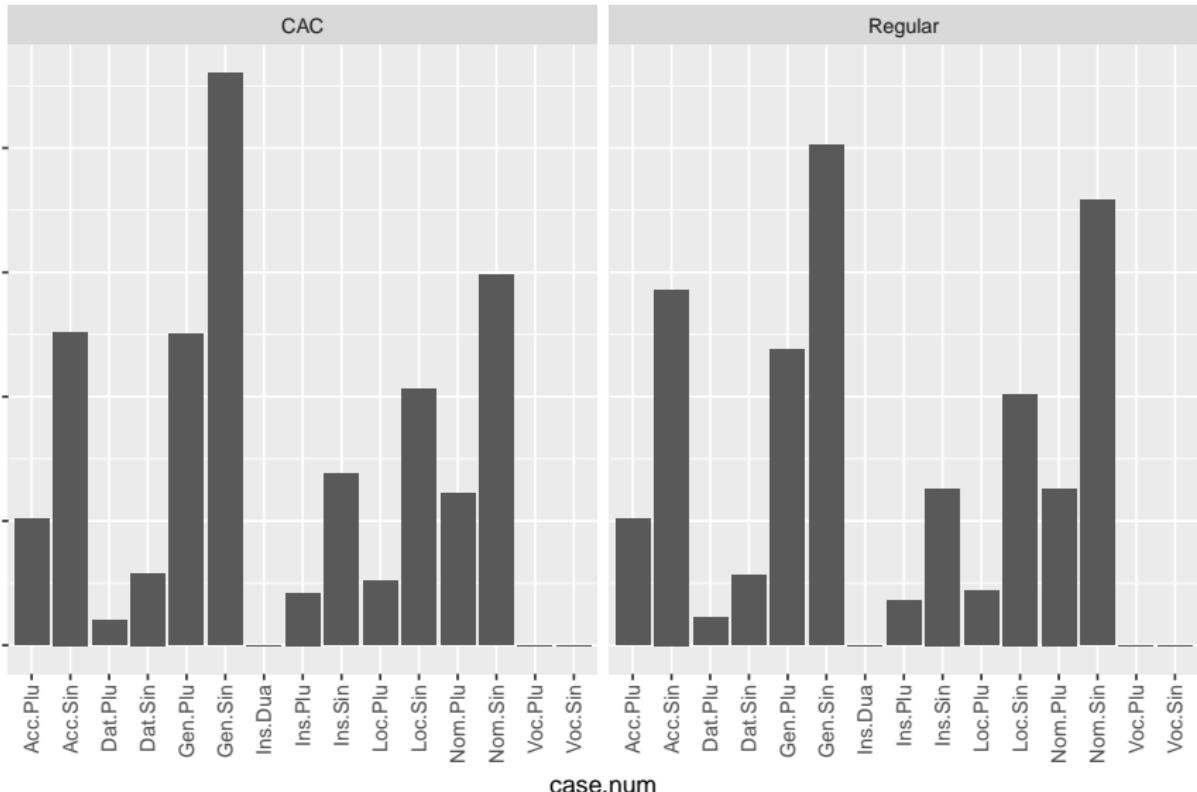
- As the corpora for the languages considered are not very homogeneous; we do not know to what extent the types of the corpora affect the use of different cases.

But: For Czech, Russian, Greek, and Latin, there are two different corpora available:

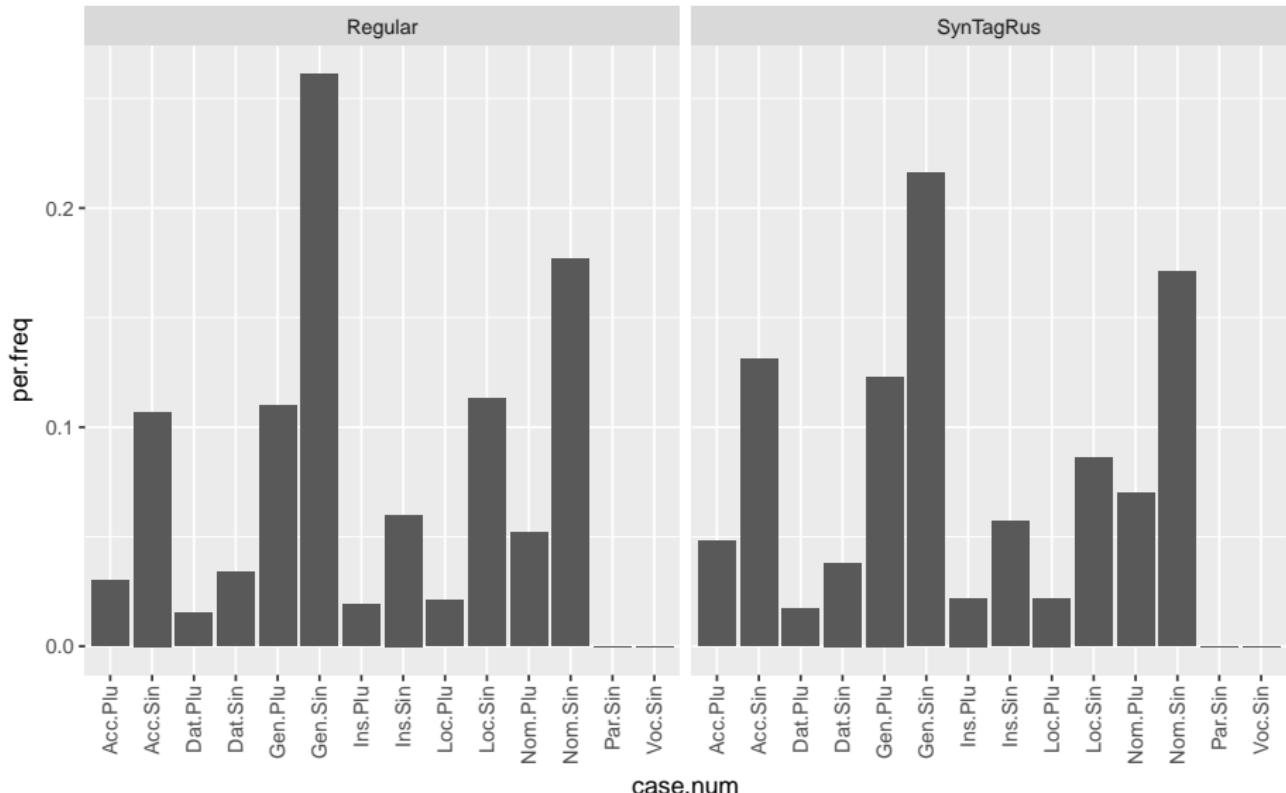
- Czech: Regular (1.3m words), CAC (482k words), overlap=0.01%
- Russian: Regular (87k words), SynTagRus (988k words), overlap=0%
- Greek: Regular (182k words), PROIEL (198k words), overlap=0.8%
- Latin: ITTB (280k words), PROIEL (159k words), overlap=0.006%

Therefore, we can compare the behaviour of different corpora for the same language in the UD data base with respect to case-number distribution.

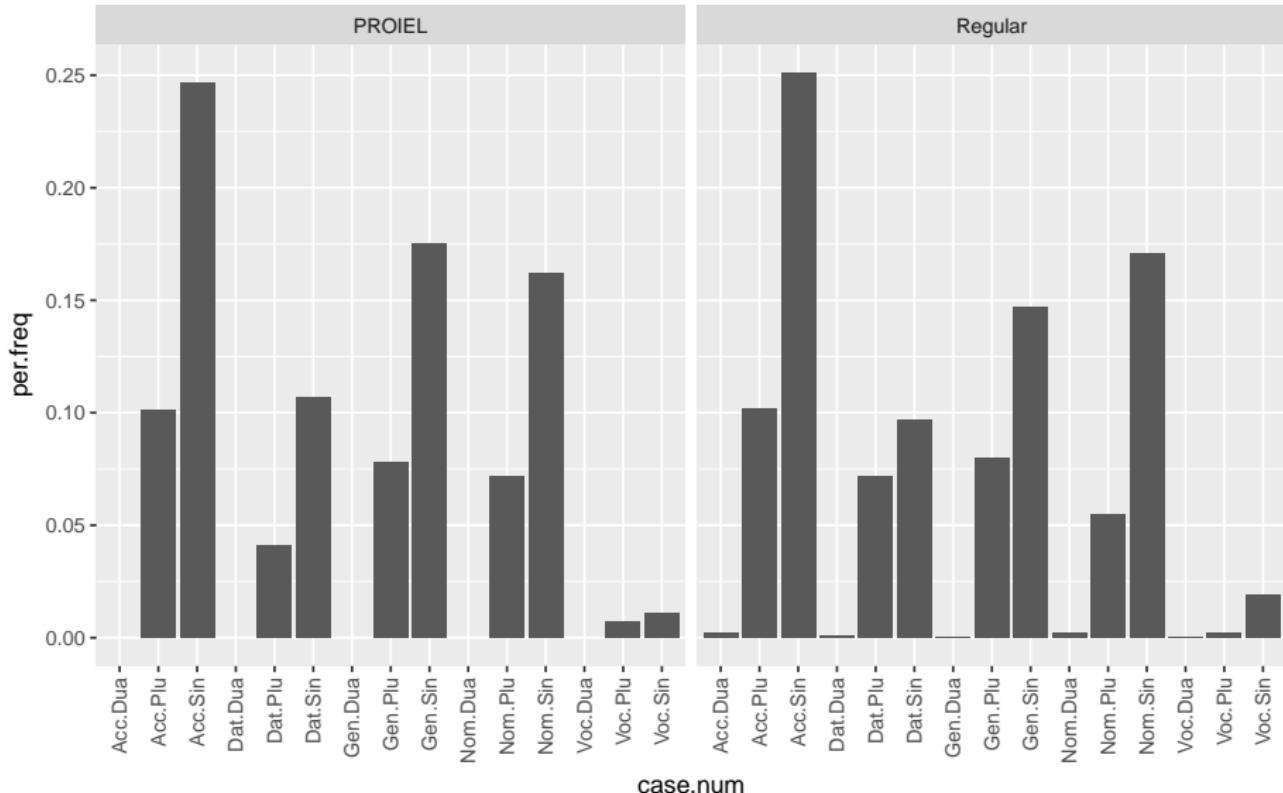
# Czech



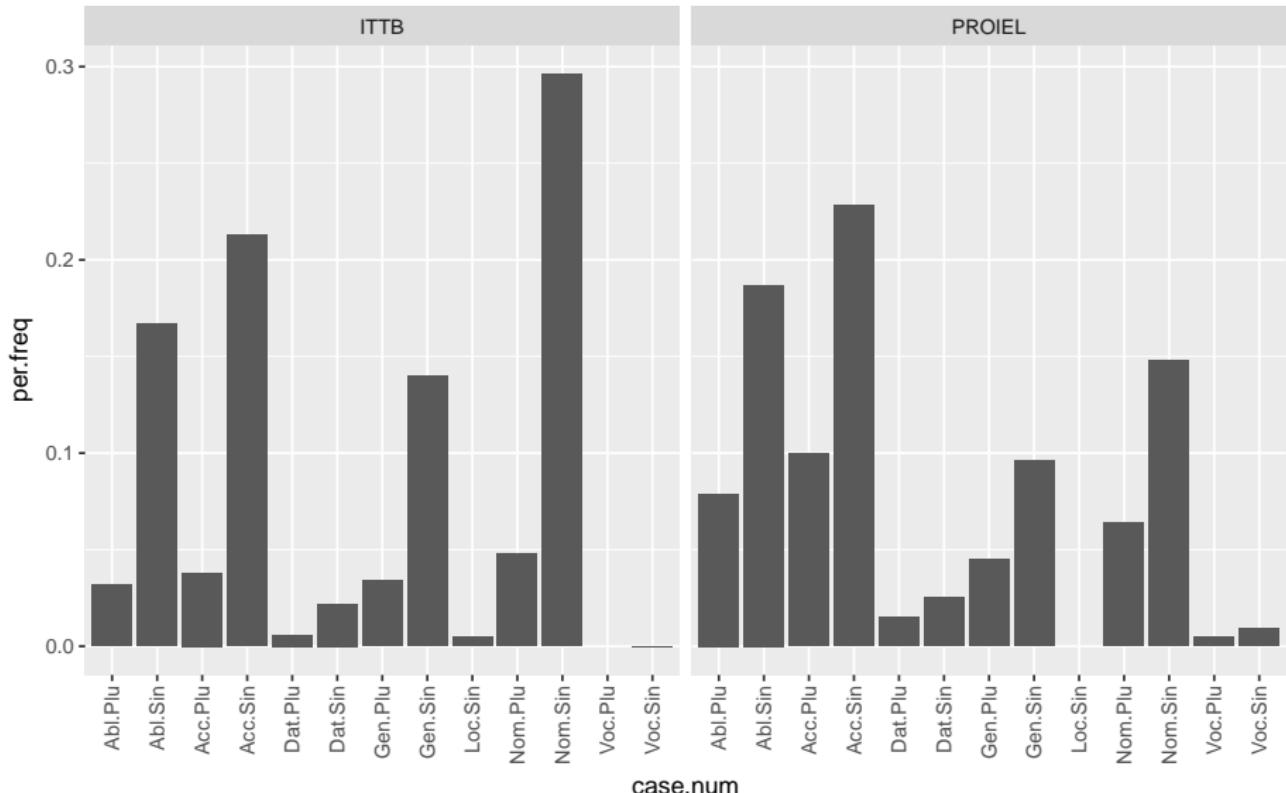
# Russian



# Greek



# Latin



Are the case frequency distributions universal?

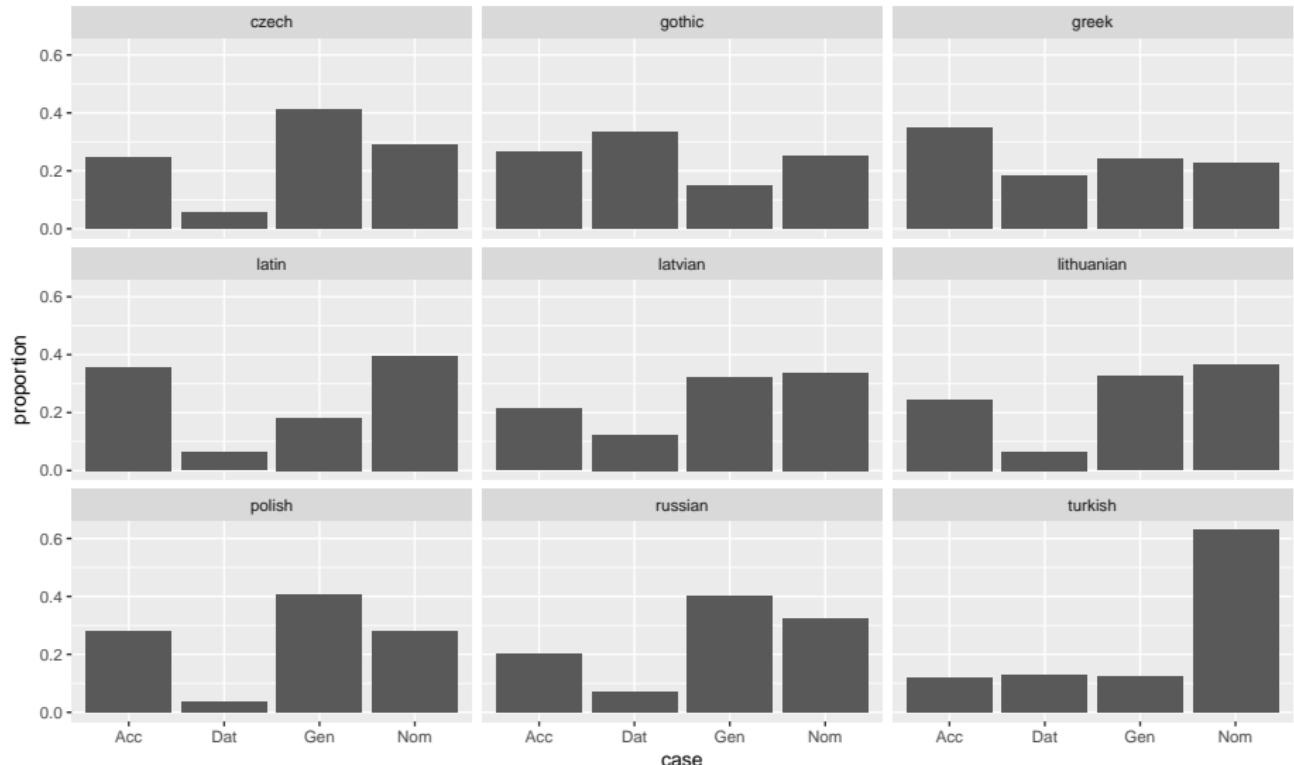
# Case frequencies

Frequency has been used in previous studies to explain coding asymmetries of case markers

However, it does not yet seem to be clear whether case frequency is language specific or stable across languages (e.g. Downing and Stiebels, 2012, p. 402)

To address this issue, we looked at the frequencies of the nominative, accusative, dative, and genitive cases in different languages for nouns and pronouns.

# Case frequencies



## Testing the FFH

## Testing the FFH

Is there a correlation between the frequency of a given case marker and its length?  
Do we find this effect across languages?

For simplicity and comparability, we only focus on the nominative, accusative, dative (except modern Greek), and genitive cases.

Length is measured as the number of segments/moras (if possible):

# Token / type frequency

We considered two kinds of frequency, token and type frequency.

In addition, the number of cells a case marker fills (i.e. the number of case values it expresses) has been noted.

The three values are calculated as follows:

- (5)     $\text{lex}_1\text{-CM (NOM.SG)}$ ;  $\text{lex}_2\text{-CM (NOM.SG)}$ ;  
       $\text{lex}_3\text{-CM (NOM.SG)}$ ;  $\text{lex}_2\text{-CM (DAT.SG)}$ ;  
       $\text{lex}_1\text{-CM (DAT.SG)}$ ;  $\text{lex}_1\text{-CM (NOM.SG)}$

**Token frequency** How often does a given marker occur in the corpus?

$$n_{\text{token}}(\text{CM})=6$$

**Type frequency** With how many different lexemes does a given marker occur in the corpus?

$$n_{\text{type}}(\text{CM})=3$$

**Number of cases** How many cells of the paradigm does a case marker fill?

$$n_{\text{cases}}(\text{CM})=2$$

# Cell distribution of the case marker

An example for the number of cases from Modern Greek:

- (6) the distribution of the marker [-is]: (Modern Greek)
- |   |        |          |
|---|--------|----------|
| $\delta\nu\pi\acute{a}\mu$ - $\varepsilon i\varsigma$ | NOM.PL | 'force'  |
| $\delta\nu\pi\acute{a}\mu$ - $\varepsilon i\varsigma$ | ACC.PL | 'force'  |
| $\mu\acute{a}\chi$ - $\eta\varsigma$                  | GEN.SG | 'battle' |

# Results. Ancient Greek

## Token frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.0000	0.1166	17.152	<2e-16	***
token.freq.scaled	-0.2502	0.1181	-2.118	0.0409	*
R-squared: 0.1082					

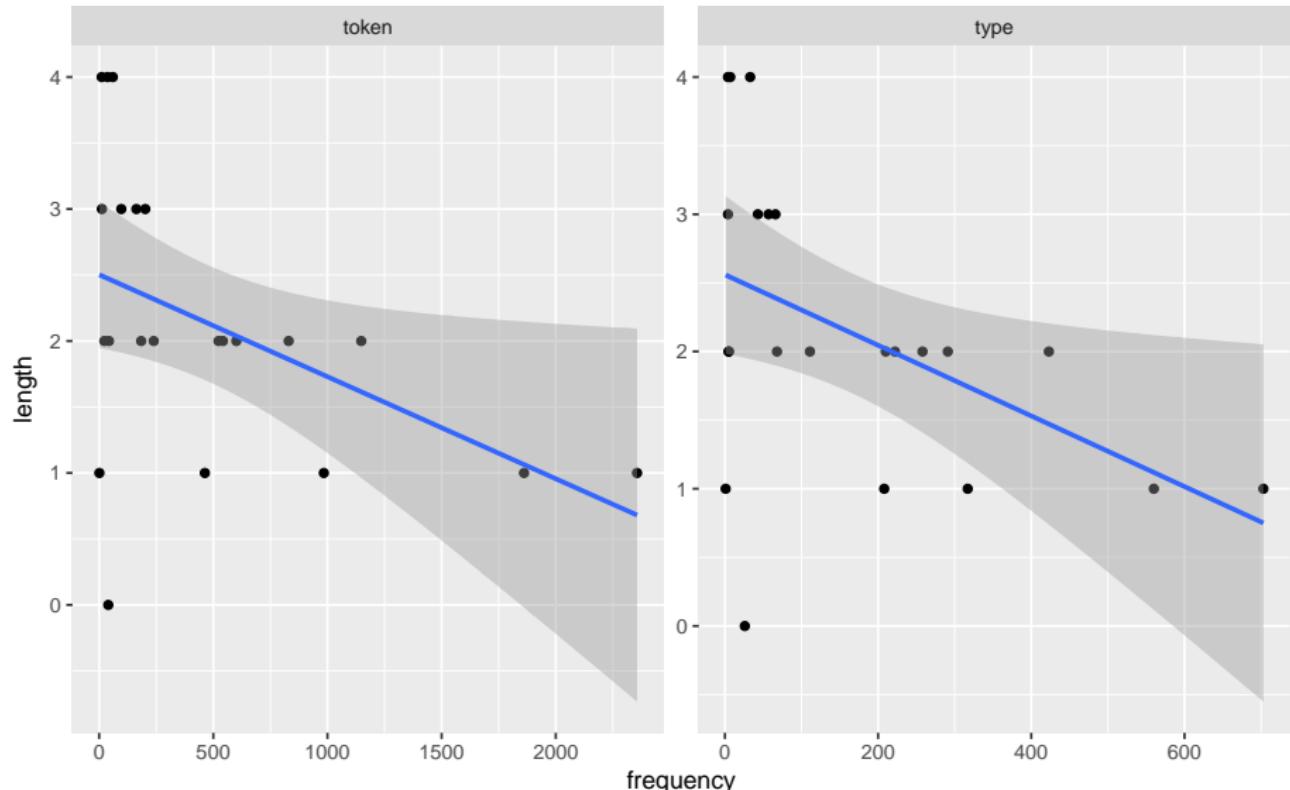
## Type frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.0000	0.1171	17.086	<2e-16	***
type.freq.scaled	-0.2421	0.1186	-2.042	0.0483	*
R-squared: 0.1013					

## Number of cases

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.0000	0.1146	17.446	<2e-16	***
ncases.numbers.scaled	-0.2826	0.1161	-2.434	0.0199	*
R-squared: 0.138					

# Results. Ancient Greek



# Results. Modern Greek

## Token frequency

Coefficients					
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.1364	0.2110	10.127	2.56e-09	***
token.freq.scaled	-0.4884	0.2159	-2.262	0.035	*
R-squared: 0.2037					

## Type frequency

Coefficients					
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.1364	0.2091	10.215	2.21e-09	***
type.freq.scaled	-0.5045	0.2141	-2.357	0.0287	*
R-squared: 0.2173					

# Results. Latin

## Token frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.7436	0.1802	15.221	< 2e-16	***
token.freq.scaled	-0.6987	0.1826	-3.827	0.000484	***
R-squared: 0.2835					

## Type frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.7436	0.1767	15.528	< 2e-16	***
type.freq.scaled	-0.6472	0.1790	-3.616	0.00091	***
type.res.ncases	-0.4598	0.2126	-2.162	0.03731	*
R-squared: 0.3302					

# Results. Gothic

## Token frequency

Coefficients					
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.7273	0.1172	23.27	< 2e-16	***
token.freq.scaled	-0.5913	0.1183	-5.00	6.65e-06	***
R-squared: 0.3205					

## Type frequency

Coefficients					
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.7273	0.1168	23.347	< 2e-16	***
type.freq.scaled	-0.5953	0.1179	-5.049	5.58e-06	***
R-squared: 0.3248					

# Results. Russian

## Token frequency

Coefficients					
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.6400	0.1314	12.484	9.99e-12	***
token.freq.scaled	-0.4932	0.1341	-3.679	0.00125	**
R-squared: 0.3704					

## Type frequency

Coefficients					
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.5775	0.3605	4.375	0.000293	***
type.freq.scaled	-0.5135	0.1284	-3.999	0.000705	***
(type.res.ncases, 3)1	1.6011	1.2647	1.266	0.220077	
(type.res.ncases, 3)2	-2.1015	0.9180	-2.289	0.033074	*
(type.res.ncases, 3)3	0.3593	0.7403	0.485	0.632676	
R-squared: 0.5081					

# Results. Polish

## Token frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.1000	0.1287	8.546	2.71e-05	***
token.freq.scaled	-0.4183	0.1357	-3.083	0.0151	*
R-squared: 0.5429					

## Type frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.1000	0.1272	8.651	2.48e-05	***
type.freq.scaled	-0.4225	0.1340	-3.152	0.0136	*
R-squared: 0.554					

# Results. Czech

## Token frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.7600	0.1236	14.234	1.40e-12	***
token.freq.scaled	-0.6458	0.1262	-5.118	3.96e-05	***
token.res.ncases	-0.3959	0.1673	-2.366	0.0272	*

R-squared: 0.591

## Type frequency

	Coefficients				
	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	1.7600	0.1224	14.377	1.15e-12	***
type.freq.scaled	-0.6561	0.1249	-5.251	2.87e-05	***
type.res.ncases	-0.3830	0.1664	-2.302	0.0312	*

R-squared: 0.5991

# To sum up

## Case frequency

- Different languages exhibit different case frequency distributions.

## The Form-Frequency Hypothesis

- Raw token frequency had some explanatory power for the length of the exponents.
  - Type frequency and the number of cases were better predictors.
  - However, frequency alone does not explain the asymmetry of the length of case markers.
- What other factors can play a role that are able to outrank the frequency effects?