

Model 1 (Python version)

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1 Model 1: A probabilistic model of segment borrowability

This report includes supplementary materials for:

Operationalizing borrowability: A case study from phonological segments

```
[6]: from collections import defaultdict, Counter
import pandas as pd

[7]: def load_cldf_dataset(path_to_values, path_to_languages):
    values = pd.read_csv(path_to_values)
    languages = pd.read_csv(path_to_languages)
    return pd.merge(left = values, right = languages, how="left",
                    left_on="Language_ID", right_on="ID")
```

A helper routine for computing typological frequencies.

When several inventories (doculects) are available for a single language, we collapse the inventories in them (i.e. take their union).

```
[8]: def get_frequencies_w_inventory_collapsing(dataset):
    glottocode_to_inventory = defaultdict(set)
    for row in dataset.itertuples():
        if not pd.isnull(row.Language_ID):
            glottocode_to_inventory[row.Language_ID].add(row.Value)
    print(f'{len(glottocode_to_inventory)} languages')
    frequencies_absolute = Counter()
    for segments in glottocode_to_inventory.values():
        for segment in segments:
            frequencies_absolute[segment] += 1
    frequencies_relative = {
        segment: count / len(glottocode_to_inventory)
        for segment, count in frequencies_absolute.items()
    }
    return frequencies_absolute, frequencies_relative
```

First, load the data from the [CLDF format](#) (Forkel et al. 2018) and combine the tables into single data frames including [PHOIBLE](#) (Moran and McCloy 2019) and [SegBo](#) (Grossman et al. 2020).

```
[9]: segbo = load_cldf_dataset('../data/segbo/cldf/values.csv',
                             '../data/segbo/cldf/languages.csv')
     phoible = load_cldf_dataset('../data/phoible/cldf/values.csv',
                                '../data/phoible/cldf/languages.csv')
```

Number of different inventories in PHOIBLE:

```
[10]: n_phoible_inventories = len(phoible.Language_ID.unique())
     n_phoible_inventories
```

```
[10]: 2177
```

Number of different inventories in SEGBO:

```
[11]: len(segbo.Language_ID.unique())
```

```
[11]: 498
```

Some languages in SEGBO are missing from PHOIBLE:

```
[12]: len(set(segbo.Language_ID) - set(phoible.Language_ID))
```

```
[12]: 199
```

We need to exclude them:

```
[13]: phoible_langs = set(phoible.Language_ID)
     segbo = segbo.loc[ segbo.Language_ID.map(lambda gltc: gltc in phoible_langs) ]
```

```
[14]: len(segbo.Language_ID.unique())
```

```
[14]: 299
```

```
[15]: len(set(segbo.Language_ID) - set(phoible.Language_ID))
```

```
[15]: 0
```

We compute borrowability factors for a segment s (b_s) following the approach by Eisen (2019). We assume that the marginal probability of borrowing of s ($P_s(\text{borrowing})$) is equal to probability of contact between a language with this segment and a language lacking this segment ($P_s(\text{contact})$) multiplied by the conditional probability of borrowing of this segment in a contact situation ($P_s(\text{borrowing}|\text{contact})$):

$$P_s(\text{borrowing}) = P_s(\text{contact})P_s(\text{borrowing}|\text{contact})$$

We approximate $P_s(\text{borrowing})$ with the empirical relative frequency of borrowing (q_s) provided by SEGBO and PHOIBLE and assume, following Eisen (2009), that $P_s(\text{contact})$ can be estimated as a product of the relative typological frequency of a segment (f_s) and its complement ($1 - f_s$):

$$P_s(\text{contact}) \propto f_s(1 - f_s)$$

Here and in Model 2, we define b_s to be $P_s(\text{borrowing}|\text{contact})$, which gives

$$q_s = f_s(1 - f_s)b_s$$

$$b_s = \frac{q_s}{f_s(1 - f_s)}$$

f_s is equal to the number of occurrences of s in PHOIBLE divided by the number of distinct languages in PHOIBLE.

q_s is equal to the number of occurrences of s as a borrowed segment in SEGBO again divided by the number of distinct languages in PHOIBLE: languages without borrowed segments were not included in SEGBO, which therefore cannot be used as a source of negative data.

```
[16]: # We cannot use vanilla relative frequencies: for SEGBO, we need the number of
      ↪ distinct languages from PHOIBLE;
      # for PHOIBLE see below.
```

```
(
    phoible_frequencies_absolute,
    -
) = get_frequencies_w_inventory_collapsing(phoible)

(
    segbo_frequencies_absolute,
    -
) = get_frequencies_w_inventory_collapsing(segbo)
```

2177 languages

299 languages

```
[17]: segbo_frequencies_relative = {
      segment: count_segbo / n_phoible_inventories
      for segment, count_segbo in segbo_frequencies_absolute.items()
    }
```

Vanilla relative frequencies from PHOIBLE produce valid results in most cases, but problems arise with some rare segments. E.g., when a rare segment was borrowed from language A to language B, it may happen that language A then quickly loses it. As a result, this segment may have a higher frequency in SEGBO than in PHOIBLE, which makes the derivation ill-defined.

In order to avoid this issue we create two versions of absolute PHOIBLE frequencies – one where the values are greater than or equal than in SEGBO and one where they are strictly greater (through Laplace smoothing) – and then use these absolute frequencies to compute relative typological frequencies.

```
[18]: phoible_greater_or_equal = {}
phoible_strictly_greater = {}
for segment, count_segbo in segbo_frequencies_absolute.items():
    if count_segbo >= phoible_frequencies_absolute[segment]:
        print(segment, count_segbo, phoible_frequencies_absolute[segment])
        phoible_greater_or_equal[segment] = count_segbo
        phoible_strictly_greater[segment] = count_segbo + 1
    else:
        phoible_greater_or_equal[segment] = phoible_frequencies_absolute[
            segment]
        phoible_strictly_greater[segment] = phoible_frequencies_absolute[
            segment] + 1
```

```
1 0
ai 1 0
1 1
ts 1 1
p 1 1
d 1 1
1 0
uə 1 1
ə 1 0
n 1 1
l 1 1
ndz 1 0
ð 1 1
```

```
[19]: phoible_freqs_relative = {
    segment: count / n_phoible_inventories
    for segment, count in phoible_greater_or_equal.items()
}
phoible_freqs_relative_laplace = {
    segment: count / n_phoible_inventories
    for segment, count in phoible_strictly_greater.items()
}
for segment, f_s in sorted(phoible_freqs_relative.items(),
                           key=lambda el: el[1], reverse=True)[:10]:
    print(f'{segment}: {f_s}, {phoible_freqs_relative_laplace[segment]}')

# Smoothing has no effect on frequent segments.
```

```
m: 0.9701423977951309, 0.9706017455213597
k: 0.9205328433624254, 0.9209921910886542
j: 0.915480018373909, 0.9159393661001378
u: 0.9150206706476803, 0.915480018373909
a: 0.9108865411116215, 0.9113458888378503
p: 0.870463941203491, 0.8709232889297198
w: 0.864951768488746, 0.8654111162149747
```

```
n: 0.8474965548920533, 0.847955902618282
t: 0.7606798346348186, 0.7611391823610473
l: 0.7266881028938906, 0.7271474506201194
```

```
[20]: # Now we can compute borrowability scores using Eisen's formula with the
      ↪normalisation constant

def borrowability_score(q_s, f_s):
    return q_s / f_s / (1 - f_s) # / 6 Ingoring the normalisation constant for
    ↪simplicity

borrowability_scores = {}
borrowability_scores_laplace = {}
for segment in segbo_frequencies_relative:
    borrowability_scores[segment] = {
        'Segment': segment,
        'Borrowability': borrowability_score (
            segbo_frequencies_relative[segment],
            phoible_freqs_relative[segment]
        ),
        'PHOIBLE_frequency_absolute': phoible_greater_or_equal[segment],
        'PHOIBLE_frequency_relative': phoible_freqs_relative[segment],
        'SEGBO_frequency_absolute': segbo_frequencies_absolute[segment],
        'SEGBO_frequency_relative': segbo_frequencies_relative[segment]
    }
    borrowability_scores_laplace[segment] = {
        'Segment': segment,
        'Borrowability': borrowability_score (
            segbo_frequencies_relative[segment],
            phoible_freqs_relative_laplace[segment]
        ),
        'PHOIBLE_frequency_absolute': phoible_strictly_greater[segment],
        'PHOIBLE_frequency_relative': phoible_freqs_relative_laplace[segment],
        'SEGBO_frequency_absolute': segbo_frequencies_absolute[segment],
        'SEGBO_frequency_relative': segbo_frequencies_relative[segment]
    }
}
```

```
[21]: borrowability_df = pd.DataFrame.from_dict(borrowability_scores).T.
      ↪sort_values(by='Borrowability', ascending=False)
```

```
[22]: # Frequently borrowed segments
borrowability_df.loc[ borrowability_df.SEGBO_frequency_absolute >= 10 ]
```

```
[22]:
```

	Segment	Borrowability	PHOIBLE_frequency_absolute	\
f	f	0.187879	968	
p	p	0.126288	1895	
		0.097833	1255	

b	b	0.081371	1385
z	z	0.081137	682
		0.078383	331
d	d	0.07082	640
v	v	0.070115	617
d	d	0.069825	1097
x	x	0.065986	411
h	h	0.064024	1258
		0.06386	782
l	l	0.055507	1582
r	r	0.053289	1099
t	t	0.052772	916
o	o	0.051489	1446
		0.049857	615
s	s	0.044023	1531
		0.043184	327
ts	ts	0.040479	519
e	e	0.040159	1482
η	η	0.024363	1424
		0.019333	846

	PHOIBLE_frequency_relative	SEGB0_frequency_absolute \
f	0.444649	101
p	0.870464	31
	0.576481	52
b	0.636197	41
z	0.313275	38
	0.152044	22
d	0.293983	32
v	0.283418	31
d	0.503904	38
x	0.188792	22
h	0.577859	34
	0.35921	32
l	0.726688	24
r	0.504823	29
t	0.420763	28
o	0.664217	25
	0.282499	22
s	0.703261	20
	0.150207	12
ts	0.238401	16
e	0.680753	19
η	0.654111	12
	0.388608	10

SEGB0_frequency_relative

f	0.046394
p	0.01424
	0.023886
b	0.018833
z	0.017455
	0.010106
d	0.014699
v	0.01424
d	0.017455
x	0.010106
h	0.015618
	0.014699
l	0.011024
r	0.013321
t	0.012862
o	0.011484
	0.010106
s	0.009187
	0.005512
ts	0.00735
e	0.008728
ŋ	0.005512
	0.004593

```
[23]: # Rare segments
borrowability_df.loc[ borrowability_df.SEGBO_frequency_absolute <= 2 ][:10]
```

```
[23]:      Segment Borrowability PHOIBLE_frequency_absolute \
ts      ts      1.00046      1
l      l      1.00046      1
p      p      1.00046      1
uə     uə     1.00046      1
ai     ai     1.00046      1
ndz   ndz     1.00046      1
      1.00046      1
n      n      1.00046      1
      1.00046      1
d      d      1.00046      1

      PHOIBLE_frequency_relative SEGBO_frequency_absolute \
ts      0.000459      1
l      0.000459      1
p      0.000459      1
uə     0.000459      1
ai     0.000459      1
ndz     0.000459      1
      0.000459      1
```

n	0.000459	1
	0.000459	1
d	0.000459	1
SEGB0_frequency_relative		
ts	0.000459	
l	0.000459	
p	0.000459	
uə	0.000459	
ai	0.000459	
ndz	0.000459	
	0.000459	
n	0.000459	
	0.000459	
d	0.000459	

```
[24]: borrowability_df.to_csv('model_1_borrowability.csv', index=False)
```

```
[25]: borrowability_laplace_df = pd.DataFrame.from_dict(borrowability_scores_laplace).
      ↪ T.sort_values(
      by='Borrowability', ascending=False)
```

```
[26]: # Frequently borrowed segments
      borrowability_laplace_df.loc[ borrowability_laplace_df.SEGB0_frequency_absolute
      ↪ >= 10 ]
```

```
[26]: Segment Borrowability PHOIBLE_frequency_absolute \
f      f      0.18784      969
p      p      0.12667      1896
      0.097862      1256
b      b      0.081415      1386
z      z      0.081072      683
      0.078189      332
d      d      0.070755      641
v      v      0.070046      618
d      d      0.069826      1098
x      x      0.065863      412
h      h      0.064043      1259
      0.063824      783
l      l      0.055565      1583
r      r      0.05329      1100
t      t      0.052757      917
o      o      0.051524      1447
      0.049808      616
s      s      0.044063      1532
      0.043075      328
ts     ts     0.040425      520
```


e	e	0.040189	1483
η	η	0.024378	1425
		0.019325	847

	PHOIBLE_frequency_relative	SEGB0_frequency_absolute \
f	0.445108	101
p	0.870923	31
	0.576941	52
b	0.636656	41
z	0.313734	38
	0.152503	22
d	0.294442	32
v	0.283877	31
d	0.504364	38
x	0.189251	22
h	0.578319	34
	0.359669	32
l	0.727147	24
r	0.505282	29
t	0.421222	28
o	0.664676	25
	0.282958	22
s	0.703721	20
	0.150666	12
ts	0.238861	16
e	0.681213	19
η	0.654571	12
	0.389068	10

	SEGB0_frequency_relative
f	0.046394
p	0.01424
	0.023886
b	0.018833
z	0.017455
	0.010106
d	0.014699
v	0.01424
d	0.017455
x	0.010106
h	0.015618
	0.014699
l	0.011024
r	0.013321
t	0.012862
o	0.011484
	0.010106

s	0.009187
	0.005512
ts	0.00735
e	0.008728
ŋ	0.005512
	0.004593

```
[27]: # Rare segments
borrowability_laplace_df.loc[ borrowability_laplace_df.SEGBO_frequency_absolute_
↳ <= 2 ][:10]
```

```
[27]:      Segment Borrowability PHOIBLE_frequency_absolute \
p      p      0.50046      2
      0.50046      2
uə      uə      0.50046      2
n      n      0.50046      2
l      l      0.50046      2
ndz  ndz      0.50046      2
ð      ð      0.50046      2
      0.50046      2
ai      ai      0.50046      2
      0.50046      2
```

	PHOIBLE_frequency_relative	SEGBO_frequency_absolute	\
p	0.000919		1
	0.000919		1
uə	0.000919		1
n	0.000919		1
l	0.000919		1
ndz	0.000919		1
ð	0.000919		1
	0.000919		1
ai	0.000919		1
	0.000919		1

	SEGBO_frequency_relative
p	0.000459
	0.000459
uə	0.000459
n	0.000459
l	0.000459
ndz	0.000459
ð	0.000459
	0.000459
ai	0.000459
	0.000459

```
[28]: borrowability_laplace_df.to_csv('model_1_borrowability_laplace.csv',  
    ↪ index=False)
```

1.1 References

Eisen, Elad. 2019. “The Typology of Phonological Segment Borrowing.” Masters thesis, Jerusalem, Israel: Hebrew University of Jerusalem.

Forkel, Robert, Johann-Mattis List, Simon J. Greenhill, Christoph Rzymiski, Sebastian Bank, Michael Cysouw, Harald Hammarström, Martin Haspelmath, Gereon A. Kaiping, and Russell D. Gray. 2018. “Cross-Linguistic Data Formats, Advancing Data Sharing and Re-Use in Comparative Linguistics.” *Scientific Data* 5: 180205.

Grossman, Eitan, Elad Eisen, Dmitry Nikolaev, and Steven Moran. 2020. “SegBo: A Database of Borrowed Sounds in the World’s Language.” In *Proceedings of the 12th Language Resources and Evaluation Conference*, 5316–22.

Moran, Steven, and Daniel McCloy, eds. 2019. *PHOIBLE 2.0*. Jena: Max Planck Institute for the Science of Human History. <https://doi.org/10.5281/zenodo.2562766>.