

The background of the slide features a repeating pattern of stylized floral and leaf motifs. The motifs are rendered in three colors: a deep purple, a light green, and a light blue. The purple motifs are the most prominent, showing a central flower-like shape with several pointed petals. The green and blue motifs are smaller, appearing as individual leaves or smaller flower buds. The overall style is clean and modern, with simple outlines and no internal shading.

# Scrambling in Persian from English Subtitles

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# Presentation Outline:

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1. Quick Description of Persian and Scrambling
2. Information on the Tehran English-Persian Parallel Corpus
3. Reorganization of Data
4. Modification of Data
5. Analysis of Data
6. Conclusions

# Part I: On Persian

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General characteristics, example  
sentence, scrambling, project motivation



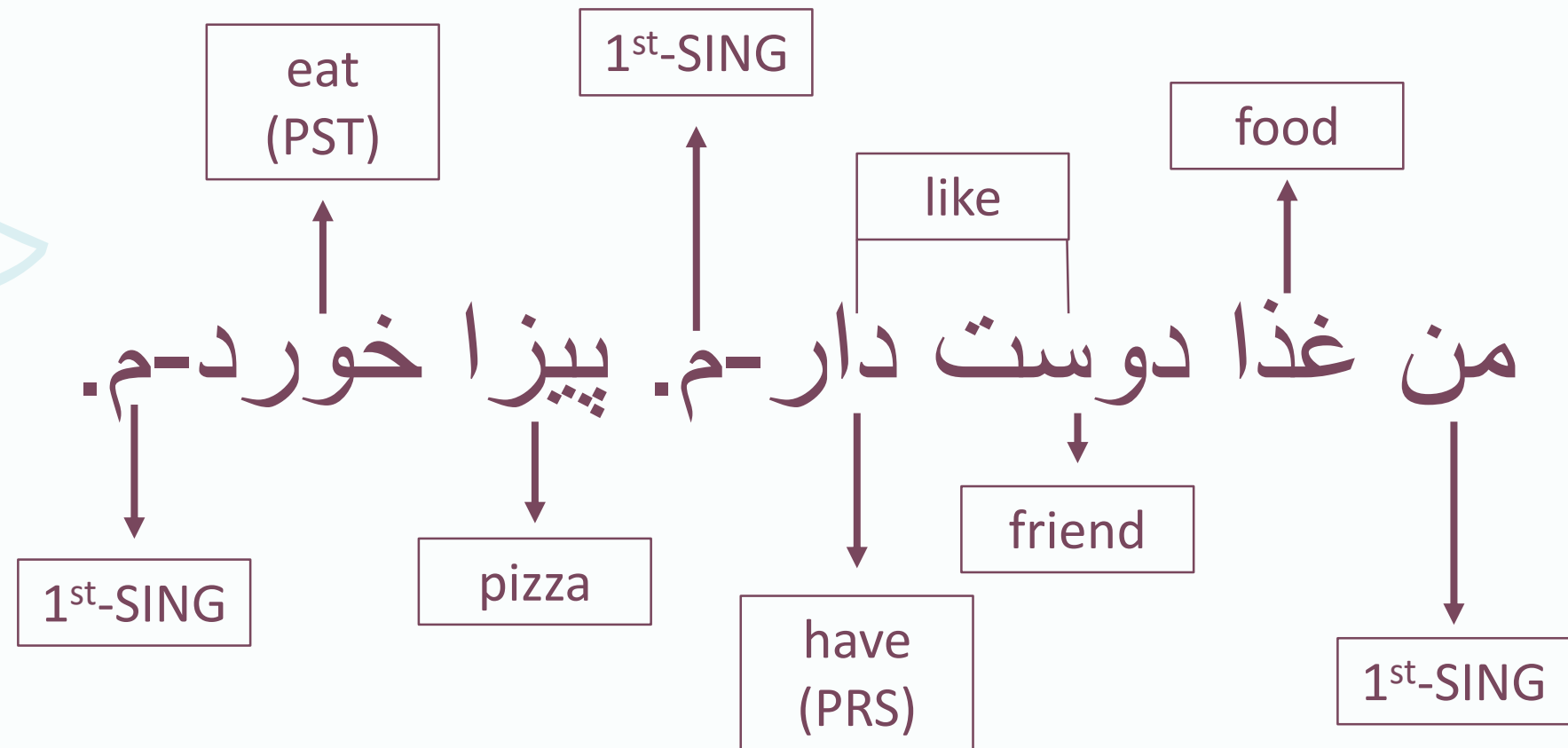
# Persian: A Brief Description

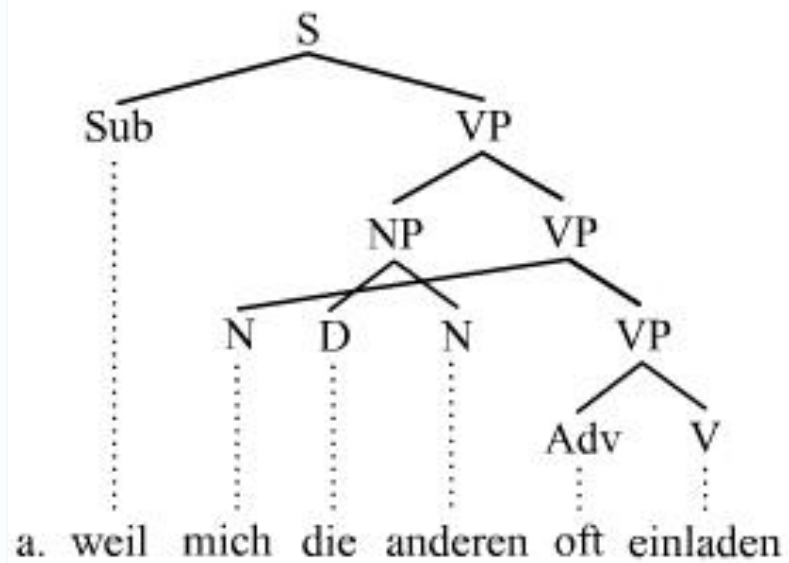
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- Indo-Iranian language spoken in Iran, Afghanistan, and Tajikistan
- Language subfamilies: Farsi (this project), Dari, and Tajik
- Reads right-to-left, using the Arabic alphabet (+4 letters)
- Many verbs are compound and include non-verbal components
- Is “pro-drop”: subject pronouns optional if already stated
- Underlyingly SOV, but is open to scrambling (DEFINE SCRAMBLING)

# An Example Sentence:

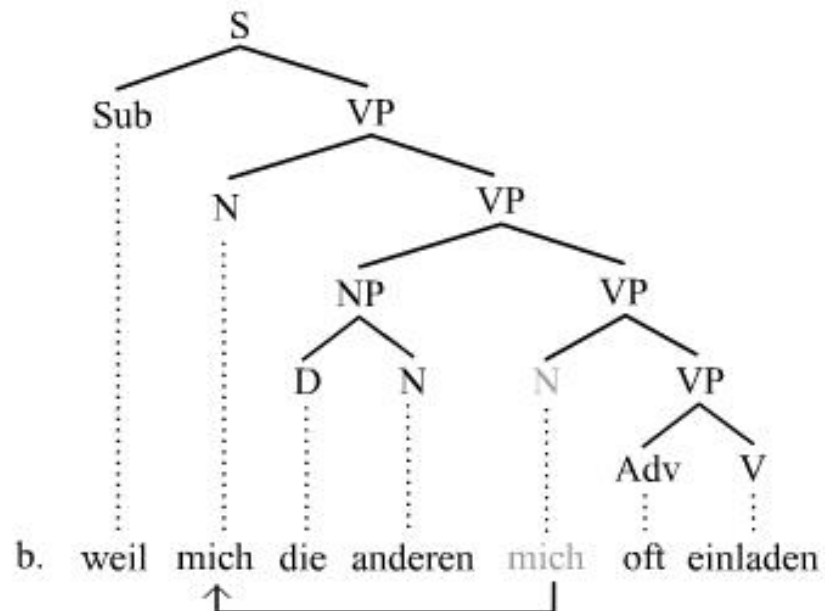
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## What is scrambling?

An example from German:  
 "Because the others invite me"





## Part II: The Tehran English- Persian Parallel Corpus

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Layout of corpus, example of raw data,  
positives/negatives of corpus



# The Tehran English-Persian Parallel Corpus: What's it all about?

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- Parallel corpus of English-Persian subtitles
- 554, 621 lines total
  - About 3.8 million words for each language
- Obtained from approx. 21,000 movie files from Open-subtitles (free online collection of movie subtitles in various languages)
- One large .xml file for each language:



# The Raw Data



```
<?xml version="1.0" encoding="utf-8"?>
<letsmt version="1.0">
<head></head>
<body>
<s id="1">raspy breathing .</s>
<s id="2">dad .</s>
<s id="3">maybe its the wind .</s>
<s id="4">no .</s>
<s id="5">stop please stop .</s>
<s id="6">you have a week , evans then well burn the house .</s>
<s id="7">william .</s>
<s id="8">god damn it , william .</s>
<s id="9">god damn it put that down .</s>
<s id="10">let go .</s>
<s id="11">its the last feed weve got .</s>
```



# The Ups and Downs of the TEP

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- Data intended to sound like regular speech, rather than formal writing
- Accurate alignment of speakers due to timestamps
- .xml files clearly indexed

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- Lack of punctuation
- Spoken Persian behaves very differently than written Persian
- Persian spelling varies greatly (but can have a great effect)
- Many incomplete sentences

# Part III:

## Organizing the Data

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Extracting data from .xml files, moving to  
DFs, combining data into one big DF

# The Process:

```
In [1]: 1 import pandas as pd
2 import numpy as np
3 import nltk
4 # import postagger
5 import xml.etree.ElementTree as ET
6 from lxml import etree
```

```
In [6]: 1 eng_lines_test = {}
2 for item in root_eng.findall('./body/s')[:5]:
3     eng_lines_test[int(str(item.values()).replac
```

```
In [7]: 1 eng_lines_test.keys()
```

```
Out[7]: dict_keys([1, 2, 3, 4, 5])
```

```
In [8]: 1 eng_lines_test.values()
```

```
Out[8]: dict_values(['raspy breathing', 'dad', 'maybe its the
```

```
In [16]: full_df = pd.Series(eng_lines).to_frame('Eng').join(pd.Series(far_lines).to_frame('Far'), how='outer')
full_df.index.name = 'ID'
full_df
```

Out[16]:

	Eng	Far
ID		
1	raspy breathing	صدای خر خر
2	dad	پدر
3	maybe its the wind	شاید صدای باد باشد
4	no	نه
5	stop please stop	دست نگه دارید خواهش میکنم دست نگه دارید

Fill & Line

```
In [21]: full_df.head()
```

Out[21]:

	Eng	Far	Eng_Tok	Far_Tok	Eng_Len	Far_Len	Eng_Types	Far_Types
ID								
1	raspy breathing	صدای خر خر	[raspy, breathing]	[صدای, خر, خر]	2	3	{breathing, raspy}	{صدای, خر}
2	dad	پدر	[dad]	[پدر]	1	1	{dad}	{پدر}
3	maybe its the wind	شاید صدای باد باشد	[maybe, its, the, wind]	[شاید, صدای, باد, باشد]	4	4	{wind, the, maybe, its}	{باشد, باید, باد, شاید, صدای}
4	no	نه	[no]	[نه]	1	1	{no}	{نه}

Fill & Line

```
In [25]: full_df.Eng_Len.value_counts()
```

```
Out[25]: 6      60403
1      59483
7      58117
5      57950
8      54154
4      52814
9      47822
10     40727
3      40255
2      36142
```

```
In [26]: full_df.Far_Len.value_counts()
```

```
Out[26]: 2      67818
5      60178
6      60088
4      56988
3      55920
7      55866
8      49803
9      42446
1      35506
10     34563
```



# Part IV: Modifying the Data

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POS-tagging, chunking, generalizing

# POS-tagging and Chunking Persian Text

```
In [0]: # Importing the necessary modules  
from hazm import *
```

```
In [3]: # Building our tagger  
tagger = POSTagger(model='postagger.model')  
tagger.tag(word_tokenize('ما بسیار کتاب می‌خوانیم'))
```

```
Out[3]: [('ما', 'PRO'), ('بسیار', 'ADV'), ('کتاب', 'N'), ('می‌خوانیم', 'V')]
```

```
In [0]: # Building our tagger  
chunker = Chunker(model='chunker.model')
```

```
In [5]: # Test file for the chunker  
tagged = tagger.tag(word_tokenize('کتاب خواندن را دوست داریم'))  
tree2brackets(chunker.parse(tagged))
```

```
Out[5]: '[VP دوست داریم] [POSTP را] [NP کتاب خواندن]'
```

# How to properly generalize the WO?

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- Data is in strings → Determine WO with regex
- Alternative methods?
- Defining NPs/VPs and defining what they're *not*:

– What is:

In [21]:

```
1 np = r'\[[^\]]+ NP\'
2 vp = r'\[[^\]]+ VP\'
```

– What isn't:

In [22]:

```
1 np_not = r'\[[^\]]+ [^N]P\'
2 vp_not = r'\[[^\]]+ [^V]P\'
```

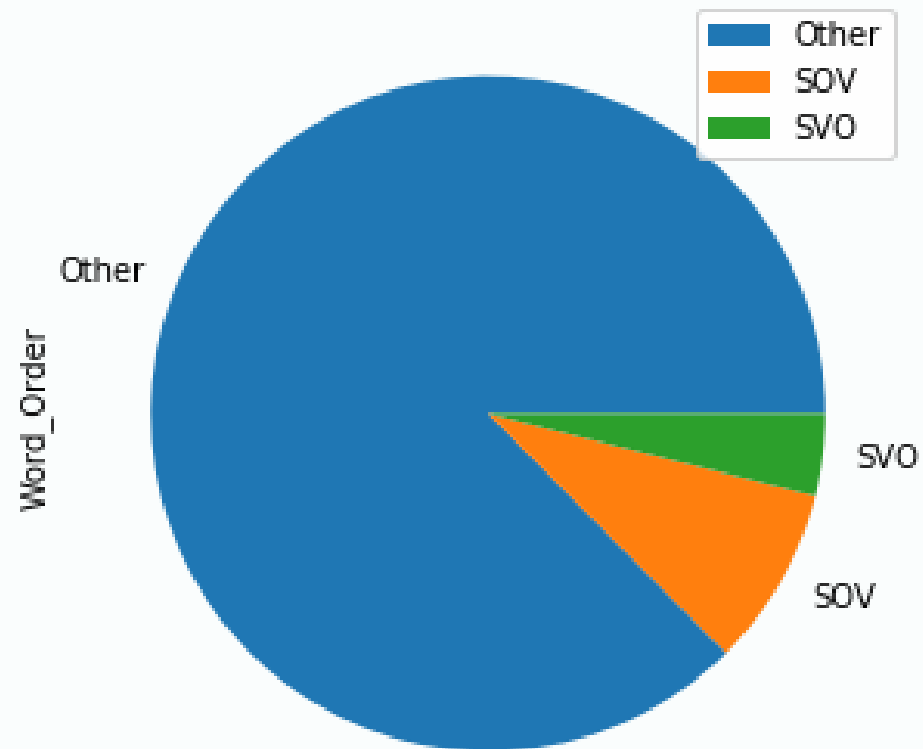


# Part V: Analyzing the Data

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How the data is split up, general remarks,  
looking at data using the word “like”





How do the  
data look?

Yikes...



# Some Qualities of Generalizations:

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- The chunker is... not great.
- SOV/SVO sentences generally correct (counts not yet discovered)
- “Other sentences” are either too simple, too complex, or have a weird quirk about them
- Looking into “is like” sentences:

# Example SOV Sentences:

```
In [36]: for item in like_sov_text[:10]:  
         print(item)  
         print()  
         print()
```

[VP است] [NP ستاره 5] [NP یک هتل] [PP مثل] [NP قلعه کاپریکورن]

[NP ميري؟] [PP داري] [ADVP كجا] [VP كني] [NP فكر مي] [PP ،باز] [NP هي آدم حقه]

[NP اين دنيا مهربوني ميبخشه] [PP به] [VP الهيه] [NP يك فرشته] [PP مثل] [NP اون زن]

[NP رنگ طلاست] [PP به] [VP خورشيد] [NP نور] [PP براي] [NP For sunlight is like gold]

[VP توش داره] [ADVP زيادي] [VP رازهاي] [NP جرم] [NP صحنه]

[NP شنه] [VP هاي] [NP دانه] [PP مثل] [NP قدرت]

# Example SVO Sentences:

In [38]:

```
1 for item in like_svo_text[:10]:  
2     print(item)  
3     print()  
4     print()
```

[NP آغاز یه مسابقه تلوزیونیه [VP شبیه [NP این ویزیتهها]

[NP مثله خوره میمونه [VP آهنگه [NP این]

[NP هنر کفش [VP هست [NP این]

[NP خلاصه میشه [NP بابا] و [NP مامان] [VP بین] [NP کورس من] [VP توی] [NP کریسمس]

[NP یک گل رز قرمزه [VP شبیه [NP عشقمون]

[VP کنی] [NP زیاد اشتباه] [VP نباید] [VP فضاست] [NP زندگی تو] [VP شبیه] [NP زندگی این پایین]

[NP پایمال کنیم] [POSTP را] [NP درست شبیه اینه که خون شهدا] [VP شرم آور] [VP های] [NP رفتار]

# What about the “Other” sentences?

In [47]:

```
1 for item in other_sample.sample(10):
2     print(item)
3     print()
```

[NP وقتی که اشراف جلسه تشکیل بدن] [PP تا]

[NP چارلی] [VP ممنونم] .

[ت] [VP دارم] [POSTP را] [NP یک همچین چیزایی] [ADVP هم] [NP من] [VP رفتم] [NP اون خونه] [PP به] [NP من] [که] [NP وقتی] [PP از] [NP جریه میکنم]

[NP اون پسر خشمگین شدن] [PP نسبت به] [NP اونها]

[NP تخم هارو داغ و خوب میکنیم اینجوری سفت و محکم میشن] [NP ما] [ADVP حالا]

[NP میکنن انجام میدی] [PP بقیه] [NP هر کاری که] [NP تو] [VP نداشتی] [NP چند ماهه وضع خوبی] [NP تو] [ADVP پس]

میبینم [VP این شکلی] [NP اونو] [NP من] و

[پیش من] [VP اون اومد] ، [VP باش] [NP هان آروم] ، [NP تولیور] ، [NP بین شما] [VP بوول بیاد] [VP گذاشتید] [NP چه علت] [PP به] [NP]

[VP می شناسند؟] [POSTP را] [NP همدیگر] [NP آنجا] [PP از] [NP مینگ جو] و [NP یانگوم] [ADVP پس]

[NP غمزده عشق]

# Part VI: Conclusions

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(Preliminary) results, further directions,  
possible corrections



# Preliminary Results

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- Constructions from similes lean towards SOV, whereas metaphorical constructions lean towards SVO
- More abstract comparisons also lean towards SVO
- Many different constructions will be analyzed



## Limitations of Project:

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- Persian's inconsistent transcription and subject availability
- Persian parser/chunker/generalizer are limited and do not appear to have high levels of accuracy
- Nearly 80% of the data is categorized as “Other”





# Further Directions:

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- Machine learning program that categorizes sentences for scrambling
- Working with the speech data itself
- Sentiment marking



# THE END

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Questions?

# References:

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- Link to Tehran-Parallel Corpus: <http://opus.nlpl.eu/TEP.php>
- Information on the TEP: <https://link.springer.com/content/pdf/10.1007%2F978-3-642-19437-5.pdf>
- M. T. Pilevar, H. Faili, and A. H. Pilevar, “TEP: Tehran English-Persian Parallel Corpus”, in proceedings of 12th International Conference on Intelligent Text Processing and Computational Linguistics (CICLing-2011).
- Information on Hazm: <https://github.com/sobhe/hazm>
- Project Github: <https://github.com/Data-Science-for-Linguists-2019/Scrambling-in-English-to-Persian-Subtitles>