Studying the Differences Between Natural and **Programming Languages** Casey Casalnuovo Prem Devanbu

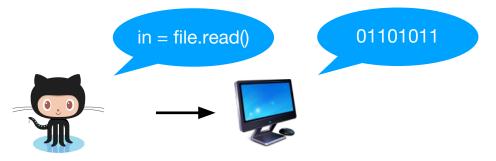


Grant #1414172 Exploiting the Naturalness of Software



in = file.read()



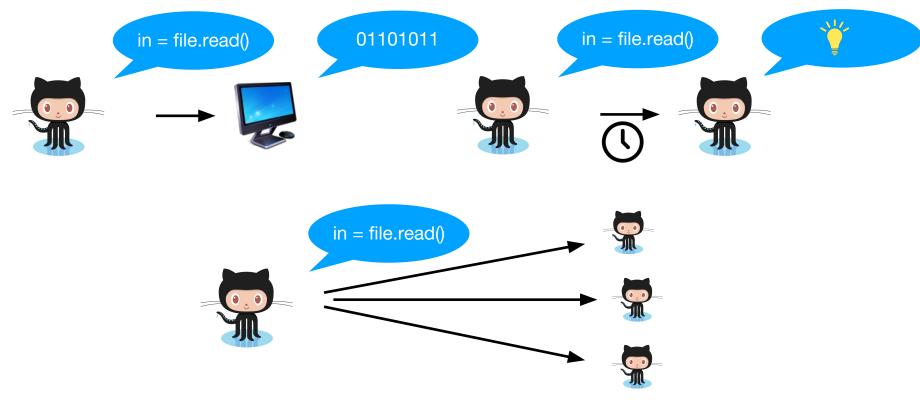












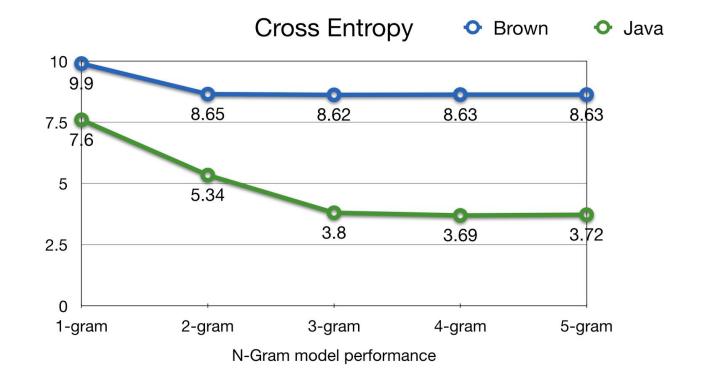


"Let us change our traditional attitude to the construction of programs: Instead of imagining that our main task is to instruct a computer what to do, let us concentrate rather on explaining to human beings what we want a computer to do."

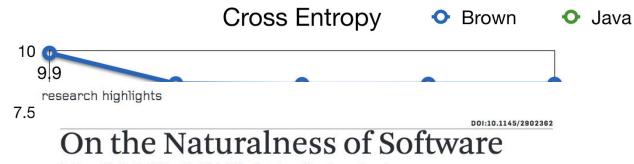
- Donald Knuth



Code Is Very Repetitive



Code Is Very Repetitive



By Abram Hindle, Earl T. Barr, Mark Gabel, Zhendong Su, and Premkumar Devanbu

Abstract

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1.

Natural languages like English are rich, complex, and powerful. The highly creative and graceful use of languages like English and Tamil, by masters like Shakespeare and Avvaiyar, can certainly delight and inspire. But in practice, given cognitive constraints and the exigencies of daily life, most human utterances are far simpler and much more repetitive and predictable. In fact, these utterances can be very usefully modeled using modern statistical methods. This fact has led to the phenomenal success of statistical approaches to speech recognition, natural language translation, questionanswering, and text mining and comprehension.

We begin with the conjecture that most software is also natural, in the sense that it is created by humans at work, with all the attendant constraints and limitations—and thus, like natural language, it is also likely to be repetitoo cumbersome to perform practical tasks at scale. Both these approaches essentially dealt with NLP from first principles—addressing *language*, in all its rich theoretical glory, rather than examining corpora of actual *utterances*, that is, what people actually write or say. In the 1980s, a fundamental shift to *corpus-based*, *statistically rigorous* methods occurred. The availability of large, on-line corpora of natural language text, including "aligned" text with translations in multiple languages,^a along with the computational muscle (CPU speed, primary and secondary storage) to estimate robust statistical models over very large data sets has led to stunning progress and widely available practical applications, such as statistical translation used by translate.google.com.^b

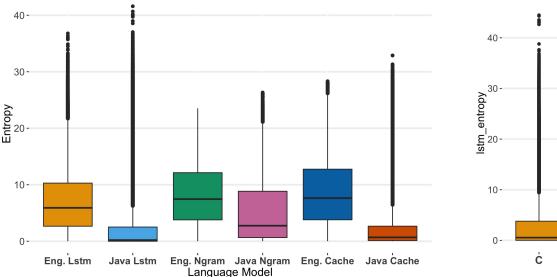
Can we apply these techniques *directly* to software, with its strange syntax, awash with punctuation, and replicate this success? The funny thing about programming is that it is as

Code is very repetitive and predictable compared to English

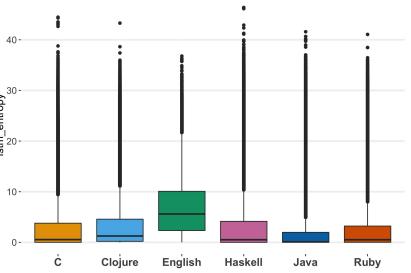
And this can be found regardless of the language model used...

Lstm, Trigram, and Trigram Cache Entropy Boxplot

Or regardless of the type of programming language selected...



Entropy Boxplot for LSTM Models



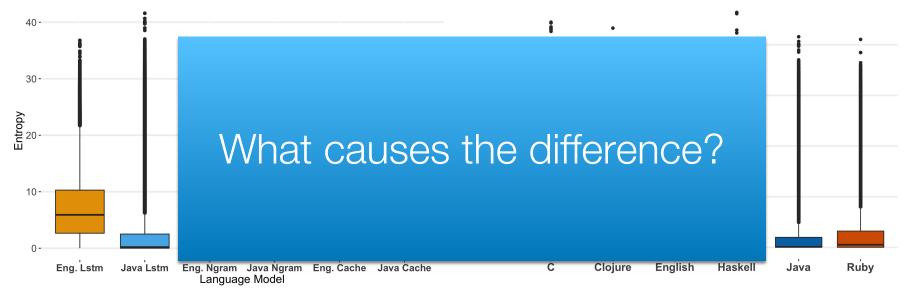
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Entropy Boxplot for LSTM Models



Two Categories of Theories

Model?

Syntactic/Structural

- Code grammar is unambiguous
- Closed category non-content words usage
- Code must compile

Social/Cognitive

- Different requirements of technical tasks
- Programming is cognitively more difficult => write more repetitive
- Community Convention

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Total Difference

???

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Open and Closed Vocabularies

Open Category = Types of words to which new elements can be added freely.

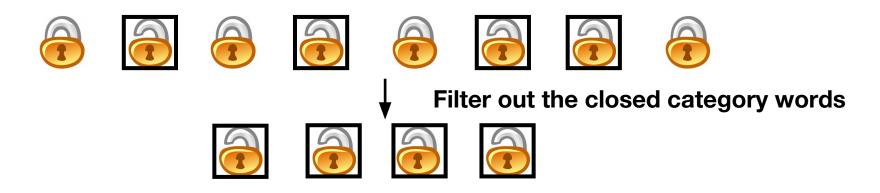
- Java = variable names, types, method and class names (literals?)
- English = nouns, verbs, adjectives, etc...



Closed Category = Types of words to which new elements cannot be added.

- Java = operators, punctuation, etc.
- English = 'stopwords' (conjunctions, pronouns, articles, etc.), punctuation.

A simple experiment



Null Hypothesis:

If the differences between Code and English are due to

- 1. Code having many more closed category words...
- 2. And these words being more predictable...

Then we would find no difference in these reduced sequences.

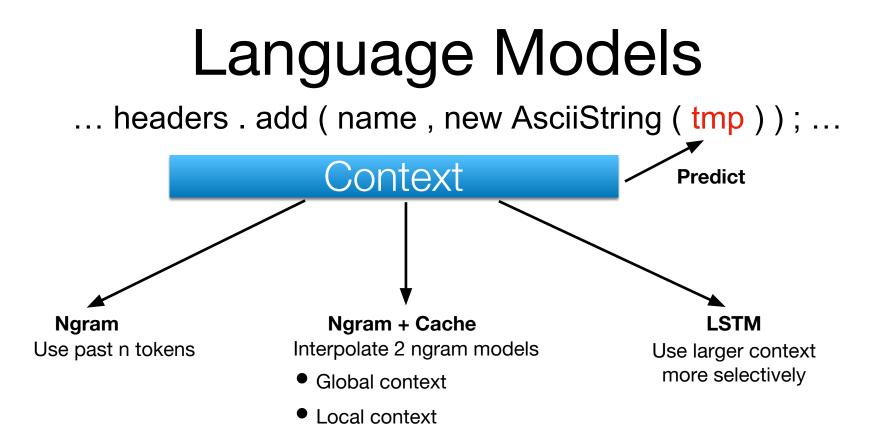
Example Open/Closed Vocabulary texts

Java

... String lines data split response setContentType response setCharacterEncoding int batchCount String s lines s s trim ...

English

... Now 175 staging centers volunteers coordinating get vote efforts said Obama Georgia spokeswoman Caroline Adelman ...

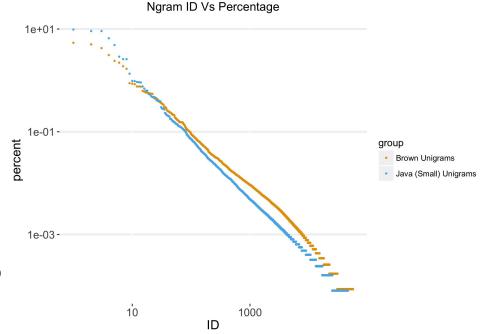


Entropy: Measure of Predictability. How many bits are necessary to represent the information? How surprising is the next token to the model?

Zipf Plots

- A commonly used plot for examining vocabulary distribution.
 - Order all tokens in decreasing order
 - Compare this rank (x-axis) against the tokens frequency.

Idea: Extend the notion of this plot from single tokens to *sequences* of tokens => provides another way to measure language repetition.



Zipf Plots (Bigrams)

Ngram ID Vs Percentage

0.100percent group **Brown Bigrams** Java (Small) Bigrams 0.001-. 1e+01 1e+03 1e+05 ID

Zipf Plots (Trigrams)

Ngram ID Vs Percentage 0.100 percent group **Brown Trigrams** Java (Small) Trigrams 0.001-1e+01 1e+03 1e+05 ID

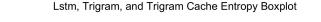
Corpora Used

	Tokens	Open Category (Ignoring Literals)	Open Category (With Literals)
English	15708917	8340284 (53.1%)	—
Java	16797357	5959414 (35.5%)	6469474 (38.5%)
Haskell	19113708	8569986 (44.8%)	10803544 (56.5%)
Ruby	17187917	3837434 (22.3%)	8992955 (52.3%)
Clojure	12553943	3283260 (26.2%)	6286549 (50.1%)
С	14172588	3707085 (26.2%)	5846097 (41.2%)

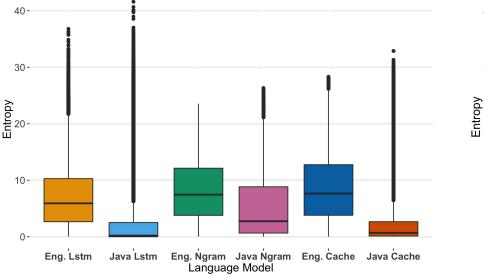
Results: Language Models

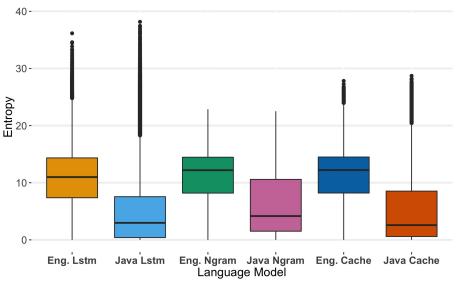
All Tokens

Open Category Tokens

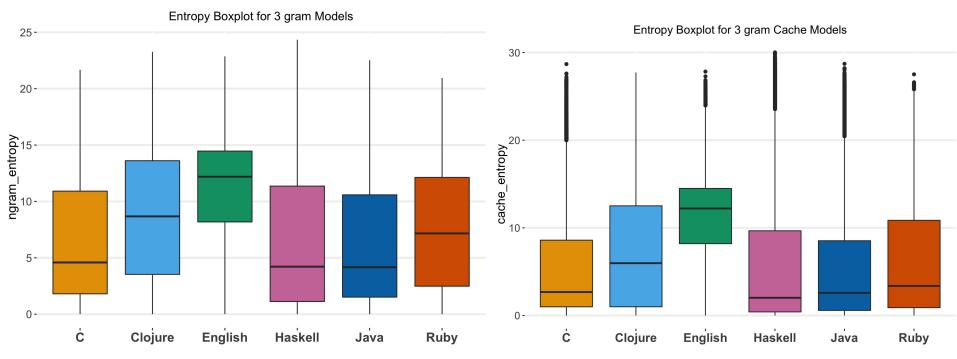


Lstm, Trigram, and Trigram Cache Entropy Boxplot



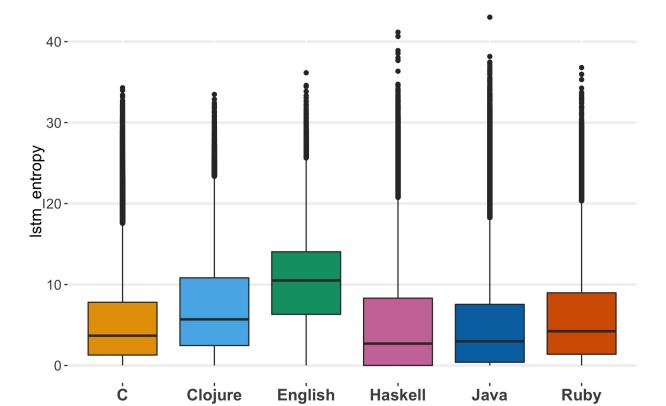


Results: Language Models



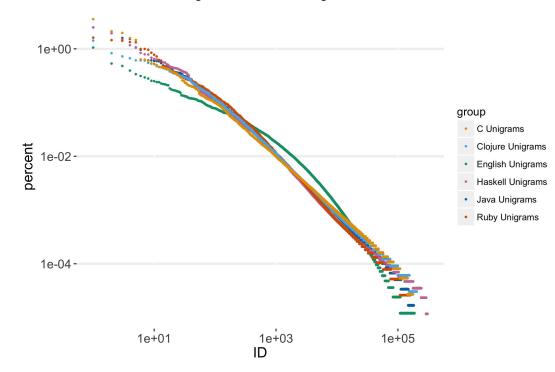
Results: Language Models

Entropy Boxplot for LSTM Models



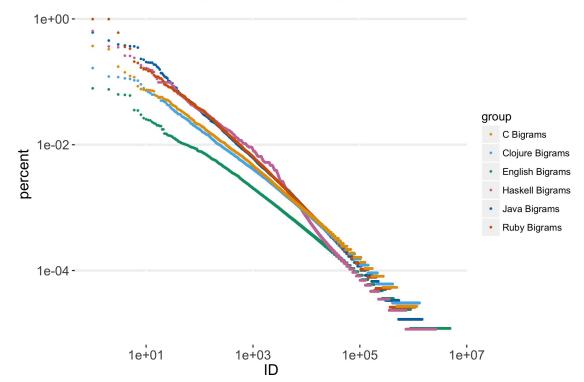
Results: Zipf Plots (Unigrams)

Ngram ID Vs Percentage



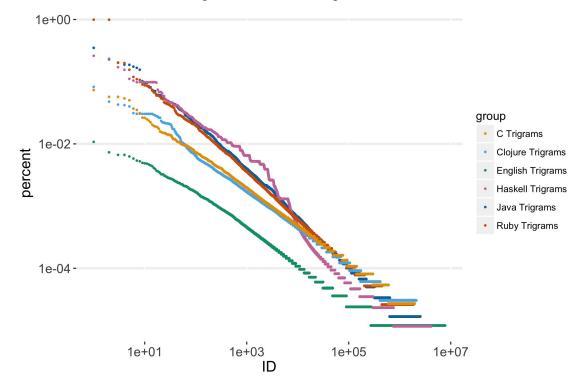
Results: Zipf Plots (Bigrams)

Ngram ID Vs Percentage



Results: Zipf Plots (Trigrams)

Ngram ID Vs Percentage



Frequent Trigram Examples

0	President Barack Obama	899
1	New York City	610
2	New York Times	556
3	George W Bush	555
4	* * *	523
5	Quote Profile Research	484
6	forward looking statements	449
7	5 per cent	438
8	NEW YORK AP	414
9	guardian co uk	406
10	published guardian co	402
11	year old man	390
12	NEW YORK Reuters	376
13	two years ago	366
14	President George W	344
15	first time since	329

0	com google common	13528
1	google common collect	8003
2	org apache cassandra	7717
3	com badlogic gdx	7286
4	int i i	5665
5	org eclipse debug	5587
6	Exception com google	3341
7	testCase com google	3335
8	org elasticsearch common	3252
9	eclipse debug internal	2589
10	java util concurrent	2526
11	org gradle api	2357
12	debug internal ui	2351
13	org apache thrift	2343
<mark>14</mark>	org nd4j linalg	2136
15	io netty handler	2063

Results Discussion

- The differences in repetition observed between English and programming languages are not merely due to the presence closed category syntactic structural words.
- In fact, the difference between them almost always increases when looking at only the open category words.

Other Experiments

- Comparing Parse Trees
 - The effect of ambiguous vs. unambiguous grammars
 - Findings: the restrictive grammar explains some, but not all of the differences.
- Comparison to Technical and English Learner Texts
 - Both display trends away from generic English and more like Code.