

# On the Hunt for Data

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# Data Annotation

# Data Annotation: who?

- Own annotators
- Volunteers
- Crowdsourcing platforms
  - *AMT, CrowdFlower*
- Expert linguists
  - *Appen, Leapforce, iSoftStone*

# Data Annotation: who?

## Crowdsourcing

- + cheap and fast
- little control over quality

## Expert Linguists

- expensive and time-consuming
- + easier to control quality

# Crowdsourcing: Amazon Mechanical Turk

- [mturk.com](https://mturk.com) - a platform for work that requires human intelligence
- Requesters vs. Workers
- Human Intelligence Task (HIT)
- Provides a sandbox: [requestersandbox.mturk.com](https://requestersandbox.mturk.com)

[Home](#)[Create](#)[Manage](#)[Developer](#)[Help](#)[New Project](#)[New Batch with an Existing Project](#)[Create HITs individually](#)

## Start a New Project

### Categorization

[Data Collection](#)[Moderation of an Image](#)[Sentiment](#)[Survey](#)[Survey Link](#)[Tagging of an Image](#)[Transcription from A/V](#)[Transcription from an Image](#)[Writing](#)[Other](#)

### Example of Categorization

#### Choose the best category for this image



- ☐ kitchen
- ☐ living
- ☐ bath
- ☐ bed
- ☐ outside

[View Instructions](#) ↓

Select the room location in home for this picture. Seating areas outside are outside not living. Offices or dens are living not bedrooms. Bedrooms should contain a bed in the picture.

You must ACCEPT the HIT before you can submit the results.

[Create Project »](#)

# Expert Linguists: Appen

- [appen.com](https://appen.com) - development of high-quality, human annotated datasets for ML
- 180 languages
- 400,000 annotators



**Appen**

@AppenGlobal

 **Follow**



Machine learning without data is like a rocket without fuel. [#DataWest17](#)

# Use case: mobile spelling corrections

- What we need?
  - Spelling error annotations in mobile phone messages.





# Use case: mobile spelling corrections

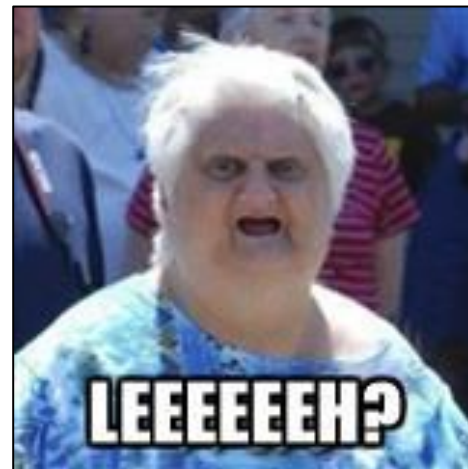
- What's available?
  - *NUS SMS Corpus*
    - 55,000 messages
  - *Mobile Forensics corpus*
    - 4,934 messages
  - *The Enron Mobile Email Dataset*
    - 2,600 messages
  - *SMS Spam Collection v. 1*
    - 425 spam messages + NUS SMS

# NUS SMS Corpus: Singlish

*Waiting in a car 4 my mum **lor**. U **leh**? Reach home already?*

*Lor* - expresses general agreeability

*Leh* - expresses negativity



The data says we need more  
data.



som<sup>ee</sup>cards  
user card

# Attack plan

- Collect data
  - Scrape Twitter
  - Use Amazon Mechanical Turk
- Annotate data
  - Automatic annotation
  - Annotation with expert linguists



# Twitter

- Twython
- 1,000 from 2011-2013 and another 1,000 from 2016
- “source” in [“iPhone”, “Android”, “Mobile”]
- Data quality:
  - language filter
  - profanities
  - too short or just hashtags
  - average word length  $< 3$ , etc.

# AMT data collection: idea 1

- What if we ask the turkers to **retype** some short messages?
  - How to set up AMT on the phone?
  - What messages to retype?
  - How do we know...
    - they are not copy-pasting?
    - they are not typing some other text instead?
    - they are using a mobile phone?
    - they are not using autocorrect?

# Results

- 10,000 sentences
- 2 days
- \$0.05 per HIT
- 33,000 misspellings

## Instructions (Click to expand)

**Important:** You must use your smartphone to complete this task. Open this task from a browser on your smartphone using the following link: [www.goo.gl/ShortLink](http://www.goo.gl/ShortLink). Type the answers using your mobile keyboard. Turn off your spell checker and autocorrect for this task. Submissions which do not use a mobile device will be rejected.

**Short link to task for smartphones:** [rebrand.ly/d7eb](http://rebrand.ly/d7eb)

If you need help turning off the spell checker, use the instructions below:

- for Android: <http://www.wikihow.com/Turn-Off-Auto-Correct-on-an-Android>
- for iOS: <http://www.howtoisolve.com/how-to-turn-off-spell-check-on-iphone-6-6-plus-ios-8-1/>

In this task, you'll be presented with 5 sentences and asked to retype the sentences as quickly as you can. Do not worry about any errors in your writing.

You will need to do the following:

- Use a mobile keyboard on your smartphone to perform the task
- Disable spellcheck / autocorrect on your phone
- Type as quickly as you can
- Do **not** go back to correct any spelling errors

# Example

*Pack my box with five dozen liquor jugs.*

*Pack my box with five dozen liquor jugs.*

*Pack my box with five dozen **liquour** jugs.*

*Pack my box with five dozen **liquir** jugs.*

***Paxk** my box with **guve** dozen **liquorr** jugs.*

*Pack my box with five dozen liquor jugs.*





## AMT data collection: idea 2

- What if we ask the turkers to ***give short answers?***
  - How to set up AMT on the phone?
  - What questions to ask?
  - How do we know...
    - they are not copy-pasting random text?
    - they are using a mobile phone?
    - they are not using autocorrect?

# Results

- 2,000 answers to 200 questions
- 4 days
- \$0.15 per HIT

## Issues:

- Misspellings cannot be extracted
- Some data bias

# Bias

*Saree. Attried in saree looks gorgeous. Its neet beautiful and sexy dress.*

*My favourite place is Guruvayur temple. I love Guruvayurappan and i feel relaxed there.*


*I live in mumbai, maharashtra, india. In mumbai there are many spots where we can enjoy...*





# Annotation


- Who: *expert linguists*
- Data: *SMS + Twitter + AMT project*
- Tool: *Anagram*


# Annotation Tool

 ANAGRAM

AnnotationProjects

Mariana Romanyshyn  
0 snippets done

Test project for mobile spelling   
13.56% complete



Save & Next Snippet

My favorite food is pizza. 8 absolutlley love it. I can have pizza at any time. I can have it cold or hot. I've rven had it for breakfast. I prefer to order it, but a homemade pizza is good too. You cannot go wrong with pizza.

Highlight text to add annotations. Click repeatedly on a token to cycle through select/insert after/insert before states.

Add correction V

8 → l

absolutlley → absolutely

rven → even

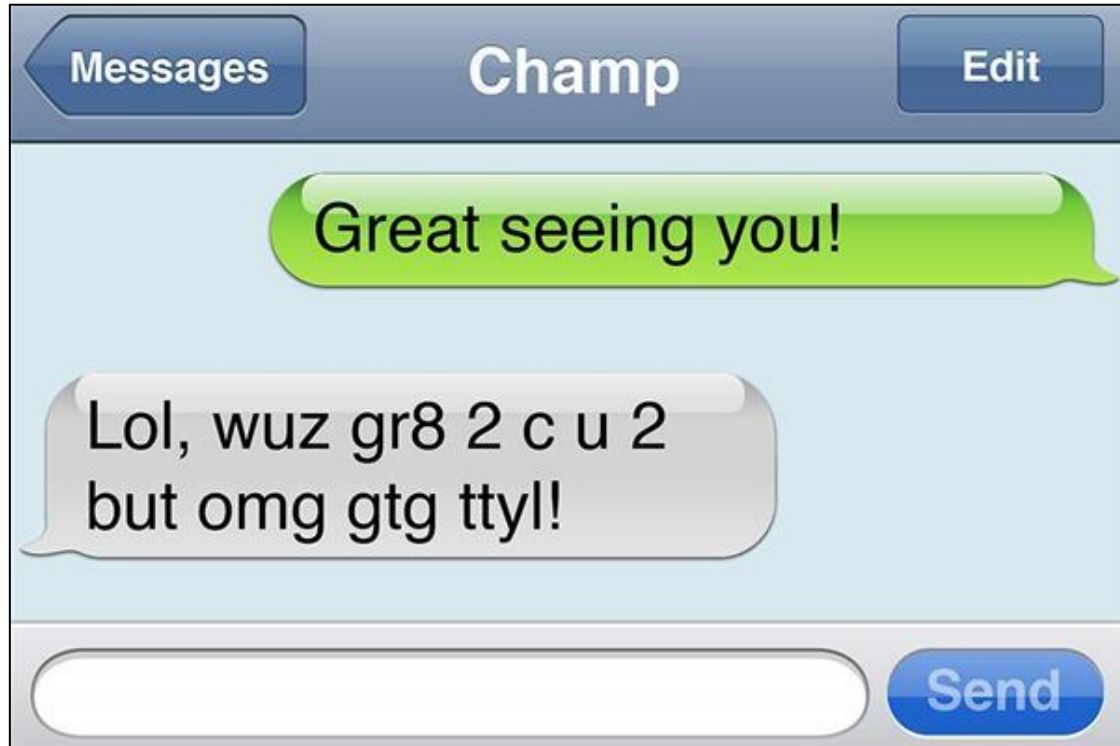
# The main issue

*could u tell ppl im gonna b a bit l8 cos 2 buses hav gon past cos they were full & im still waitin 4 1. Pete x*

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*cud u tell ppl im gona b a bit l8 cos 2 buses hav gon past cos they were full & im still waitin 4 1. Pete x*

# The main issue





# Annotation Process

- Guidelines
- Training
- Calibration
- Annotation
- Disagreement resolution

# **Data Annotation: learnings**

# Learnings

1. Guidelines
  - a. simple, short, non-contradicting
  - b. a fall-back option
  - c. as many examples as possible

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1. Guidelines
  - a. simple, short, non-contradicting
  - b. a fall-back option
  - c. as many examples as possible
2. Quality control
  - a. qualification tests / training stage
  - b. annotators with specific qualifications
  - c. cross-annotation
  - d. automatic dismissal of the work



# Learnings

3. Automatically annotated data saves time...  
*(and teach the annotators)*

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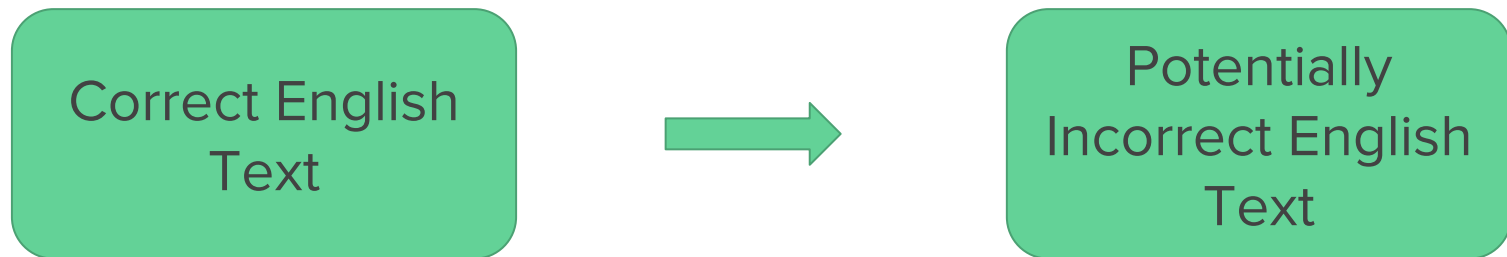


# Learnings

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4. Saving time and money
  - a. extract 100% agreement from crowdsourcing
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5. Pay quickly and be responsive to emails
6. Gamification
7. Annotation bias

# Data Generation

# The Idea



## Use case: collocation correction

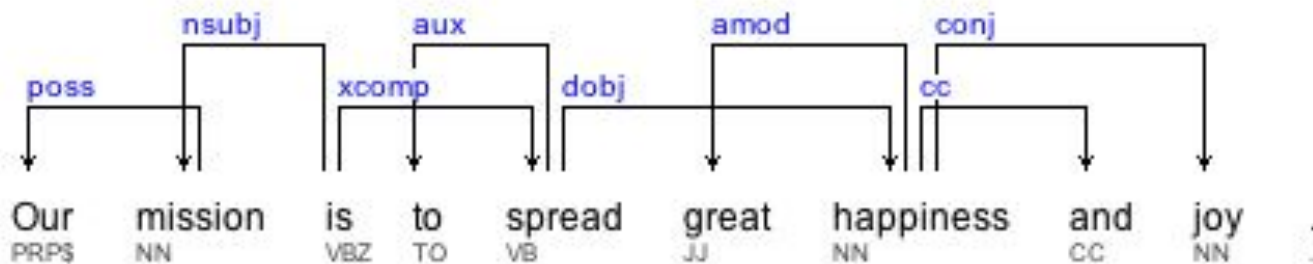
*Our mission is to spread big happiness and joy.*

## Use case: collocation correction

*Our mission is to spread {big => great} happiness and joy.*

# Use case: collocation correction

- Extract collocations from good texts



## Use case: collocation correction

- Extract collocations from good texts
- Get synonyms from a thesaurus

*Our mission is to spread **great** happiness and joy.*

*large*

*hefty*

*massive*

*big*

*considerable*

*...*

## Use case: collocation correction

- Extract collocations from good texts
- Get synonyms from a thesaurus
- Filters:
  - is the replacement a good collocation?
  - is the combination frequent?



## Use case: collocation correction

- Extract collocations from good texts
- Get synonyms from a thesaurus
- Filters
- Replace the good adjective with a synonym

*Our mission is to spread **large** happiness and joy.*

*Our mission is to spread **hefty** happiness and joy.*

*Our mission is to spread **massive** happiness and joy.*

*Our mission is to spread **big** happiness and joy.*

*Our mission is to spread **considerable** happiness and joy.*

# Results

- True positives
  - *I thought you did a {full => comprehensive} research...*
  - *...the most {beautiful => good-looking} men in the world.*
- Problems
  - Not all confusions are synonymous:
    - {crowded => heavy} traffic
  - Rare combinations can be treated as a mistake
    - {Subversive=>Underground} lines characterize...

# Data Licensing

# Data Owners

- Universities/companies/individuals
- Issues:
  - data owners have no idea about cost and/or license
  - legislation is different in different countries
  - be ready to spend about 3 months
  - and sometimes...



**JOEY DOESN'T SHARE FOOD!**

KOFLBOT

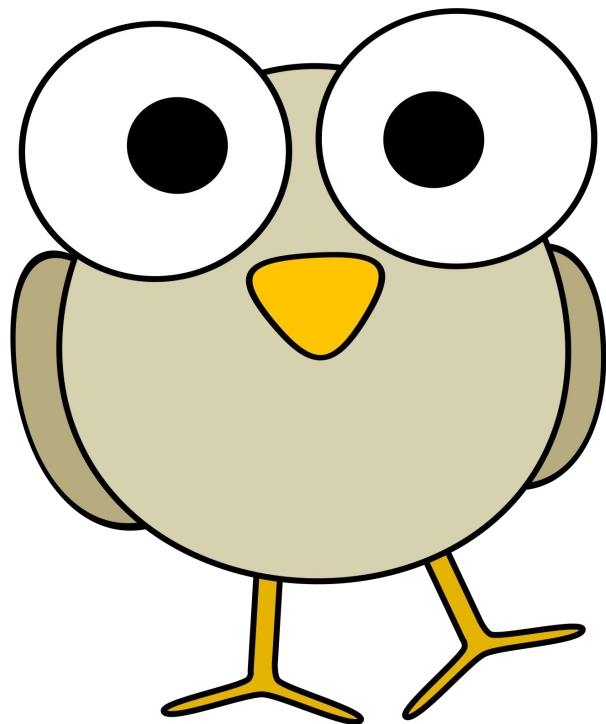
# Sometimes you win, sometimes you learn

- The Story of a Missing Licence from Creators
- The Story of a Lost Electronic Copy
- The Story of a Never-Ending Divorce
- The Story of a Grumpy Data Owner



# Conclusions

- Available data is scarce
- Data can be annotated, generated, and licenced effectively
- Good data == great effort
- Sometimes you win, sometimes you learn



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