

Fonduer: Knowledge Base Construction from Richly Formatted Data

Sen Wu

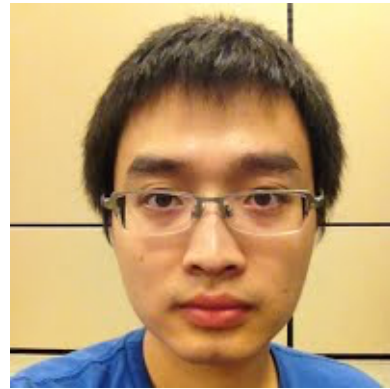
Stanford University

Thank you!

My Amazing Team:



Luke Hsiao



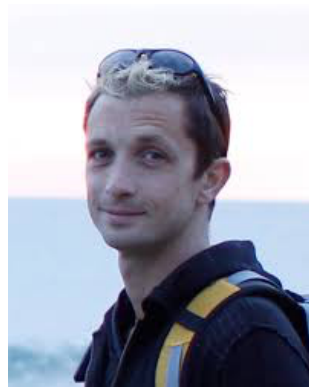
Xiao Cheng



Braden Hancock



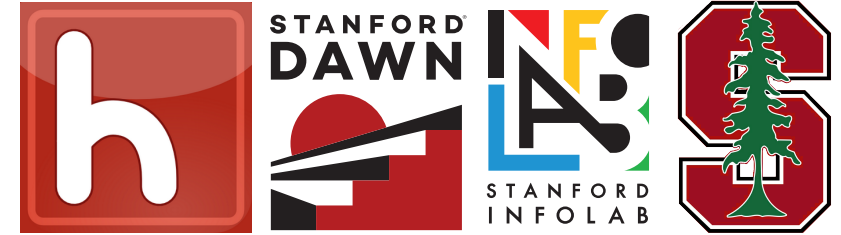
Theodoros Rekatsinas



Philip Levis



Christopher Ré



My Awesome Collaborators:



Knowledge bases are everywhere...

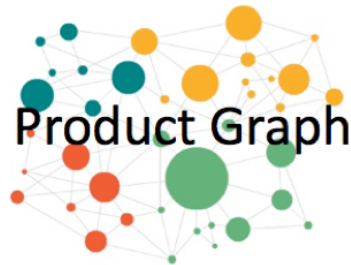
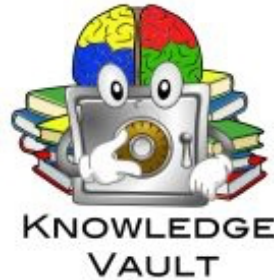


Knowledge Base Construction



Unstructured Information

Structured Knowledge Base



And many more...

But, troves of "richly formatted" information remains untapped

Richly formatted data

Richly formatted data: information is expressed via textual, structural, tabular, and visual cues.



Transistor Datasheet (PDF)

SMBT3904...MMBT3904

NPN Silicon Switching Transistors

- High DC current gain: 0.1 mA to 100 mA
- Low collector-emitter saturation voltage

Maximum Ratings

Parameter	Symbol	Value	Unit
Collector-emitter voltage	V_{CEO}	40	V
Collector-base voltage	V_{CBO}	60	
Emitter-base voltage	V_{EBO}	6	
Collector current	I_C	200	mA
Total power dissipation	P_{tot}		mW
$T_S \leq 71^\circ\text{C}$		330	
$T_S \leq 115^\circ\text{C}$		250	
Junction temperature	T_j	150	$^\circ\text{C}$
Storage temperature	T_{stg}	-65 ... 150	

Knowledge base construction from richly formatted data

Goal: extract maximum collector current from transistor datasheets

Transistor Datasheet

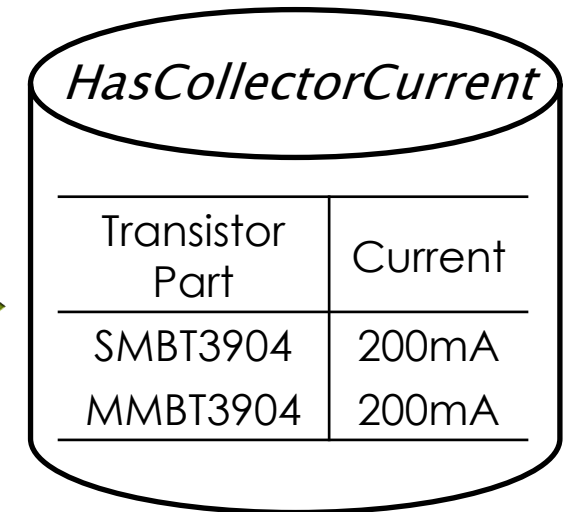
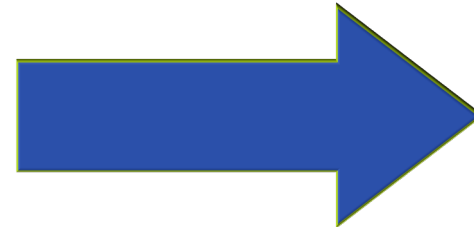
SMBT3904..MMBT3904

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Collector current	I_C	200	mA
Total power dissipation	P_{tot}		mV
$T_S \leq 71^\circ\text{C}$		330	
$T_S \leq 115^\circ\text{C}$		250	
Junction temperature	T_i	150	$^\circ\text{C}$
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Knowledge Base

Knowledge base construction from richly formatted data

Transistor Datasheet

Font: Arial, Size: 9pt, Style: Header, SMT3904, MMBT3904

NPN Silicon Switching Transistors

NPN Silicon Switching Transistors

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Emitter-base voltage	V_{EBO}	6	V
Collector current	I_C	200	mA
Total power dissipation	P_{tot}	330	mW
Junction temperature	T_j	150	°C
Storage temperature	T_{stg}	-65... 150	°C

Annotations: 'Value'; Row: 2; Column: 3; Aligned; Numbers

In richly formatted data, semantics are expressed in **textual**, **structural**, **tabular**, and **visual** modalities throughout a document

Conventional approach 1: Filter out other modalities besides unstructured text

Conventional approach 2: Limit the context scope to sentences or tables.

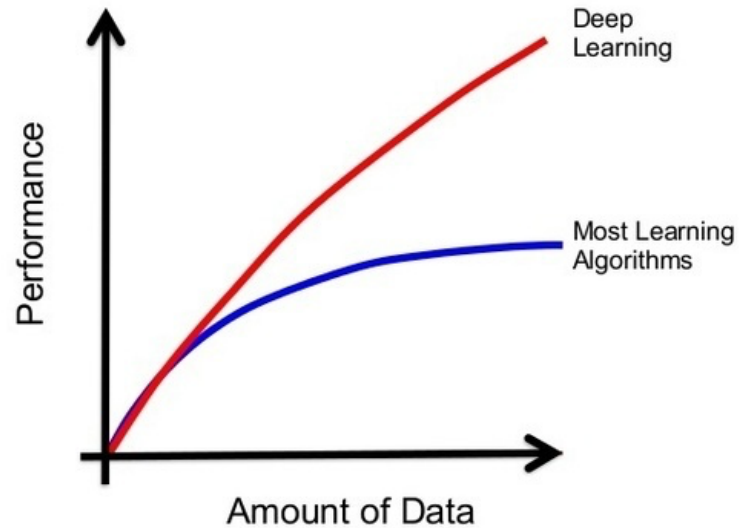
Problem: Misses important relations if you neglect multimodal information

Up to 97%
missed relations!

Can we take advantage of this powerful tool and apply it to our problem?

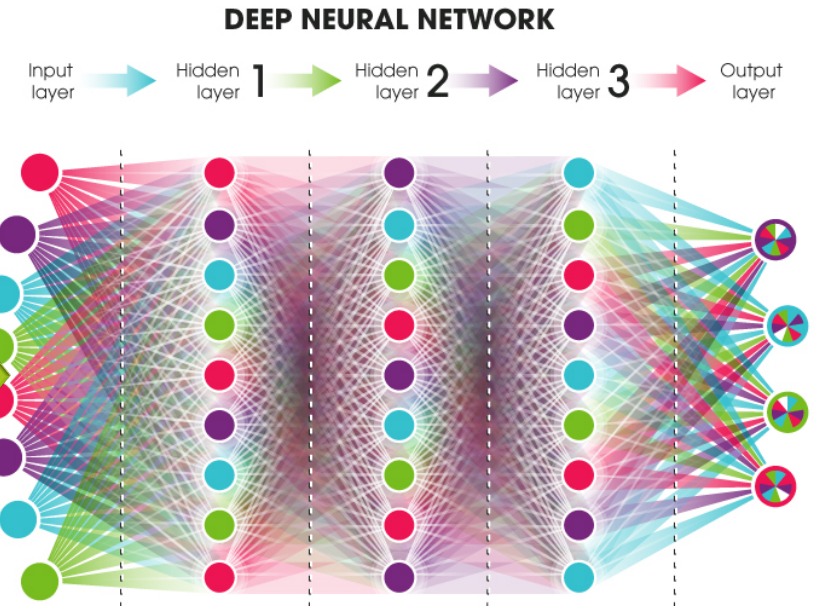
Keys to utilizing deep learning

BIG DATA & DEEP LEARNING



How do we gather enough labeled, richly formatted data?

SMBT3904...MMBT3904			
NPN Silicon Switching Transistors			
• High DC current gain: 0.1 mA to 100 mA			
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Maximum Ratings			
Parameter	Symbol	Value	Unit
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neuralnetworksanddeeplearning.com - Michael Nielsen, Yoshua Bengio, Ian Goodfellow, and Aaron Courville, 2016.

How do we model the characteristics of richly formatted data in DL?



Fonduer

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A weakly supervised deep learning framework for
knowledge base construction from richly formatted data

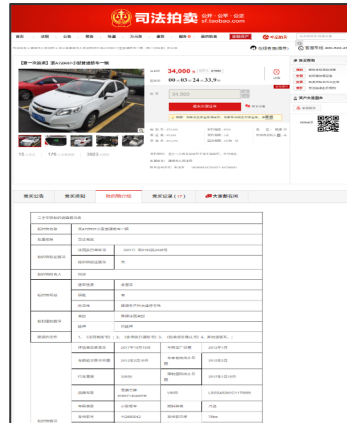
Fonduer in practice!



accenture



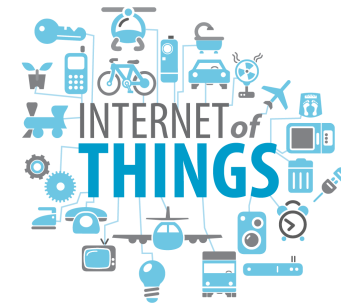
Anti-Human Trafficking



Search Engine



Genome-wide
Association Studies



Internet
of Things



Paleontology

Fonduer pipeline



FONDUER

SMBT3904_MMBT3904
SMBT3904...MMBT3904
SMBT3904...MMBT3904

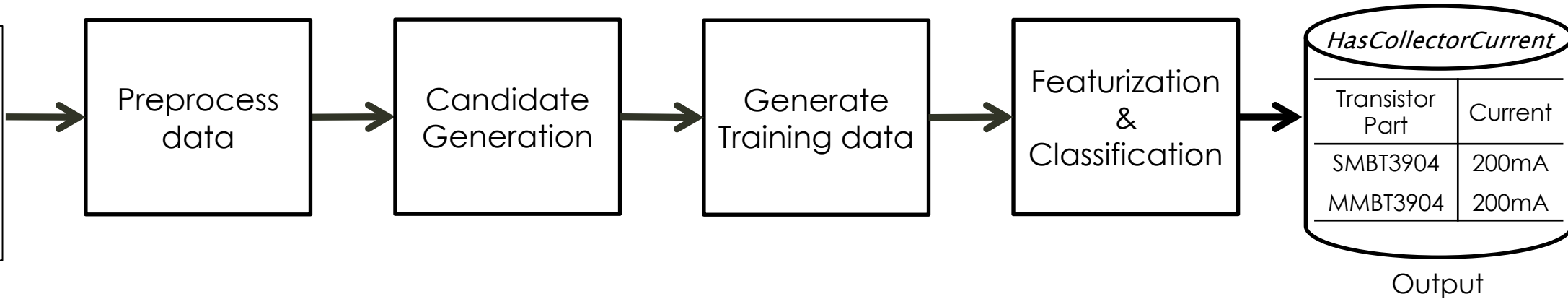
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Data Input



Generating richly formatted training data

Multimodal weak supervision

Transistor Datasheet

SMBT3904..MMBT3904			
NPN Silicon Transistors • High DC current gain: 0.1 mA to 100 mA • Low collector-emitter saturation voltage			
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Doc. level Candidates	Supervision	
	Manual	Labeling function
SMBT3904	100	✗
MMBT3904	200	✓

Weak supervision: express any supervision signal via labeling functions to generate training data

```
# Check if current is in the same row with keyword `collector`
def in_the_same_row_with(candidate):
    if 'collector' in row_ngrams(candidate.current):
        return 1
    else: return -1
```

Modeling Weak Supervision



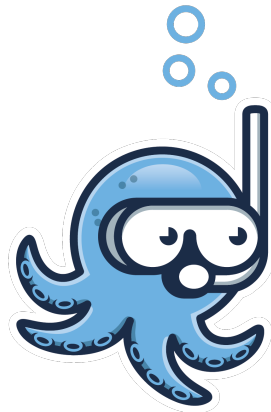
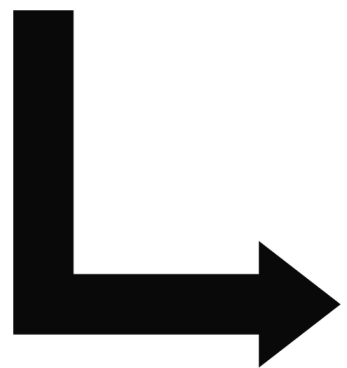
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Doc. level Candidates		Multimodal Supervision		
		Vertically aligned with 'Value'	Row ngrams contain 'mA'	'current' in sentence
SMBT3904	100	✗	∅	✓
SMBT3904	200	✓	✓	✗
SMBT3904	150	✓	✗	✗

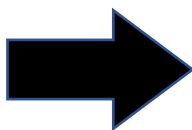
∅ = Abstain

Intuition: Use agreements / disagreements to learn the accuracy of LFs without ground truth

Output: Probabilistic Training Labels



Data programming/MeTal

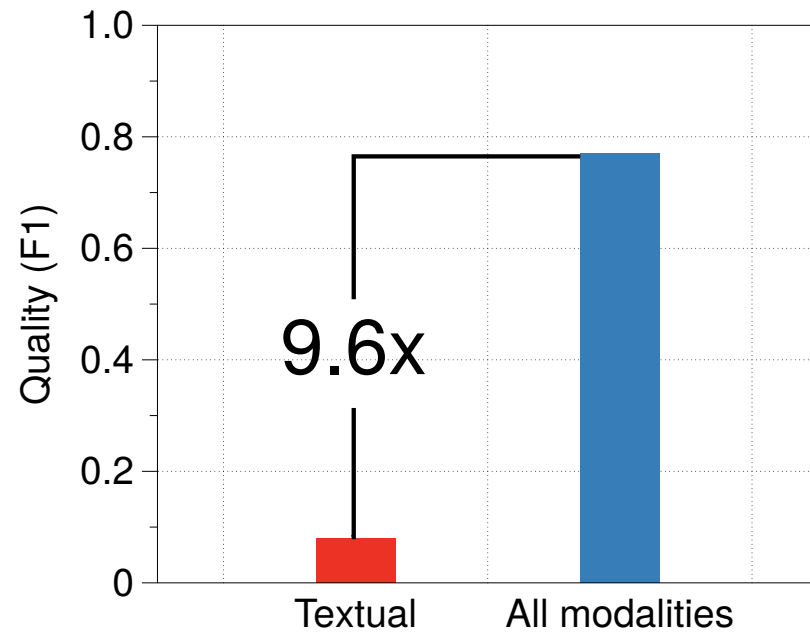


SMBT3094	100	0.5
SMBT3094	200	0.85
SMBT3094	150	0.15

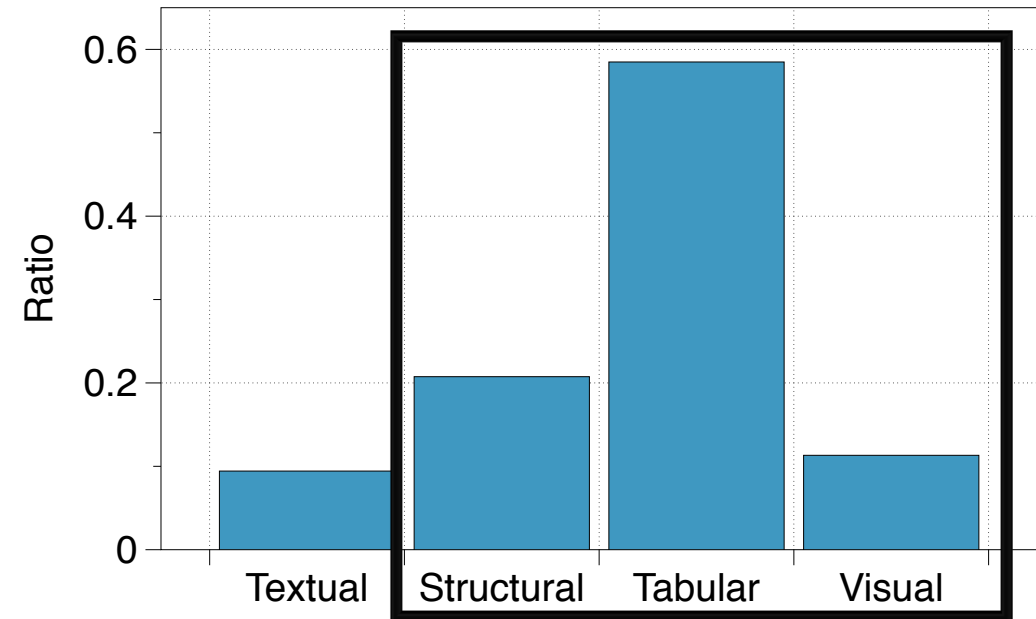
Multimodal supervision is key to quality

For transistor datasheets...

Different supervision resources' effect



Modality distribution of supervision



Users intuitively rely on multimodal information for supervision

Featurization and Classification for Richly Formatted Data

LSTM for Textual Information



Transistor Datasheet

SMBT3904...MMBT3904

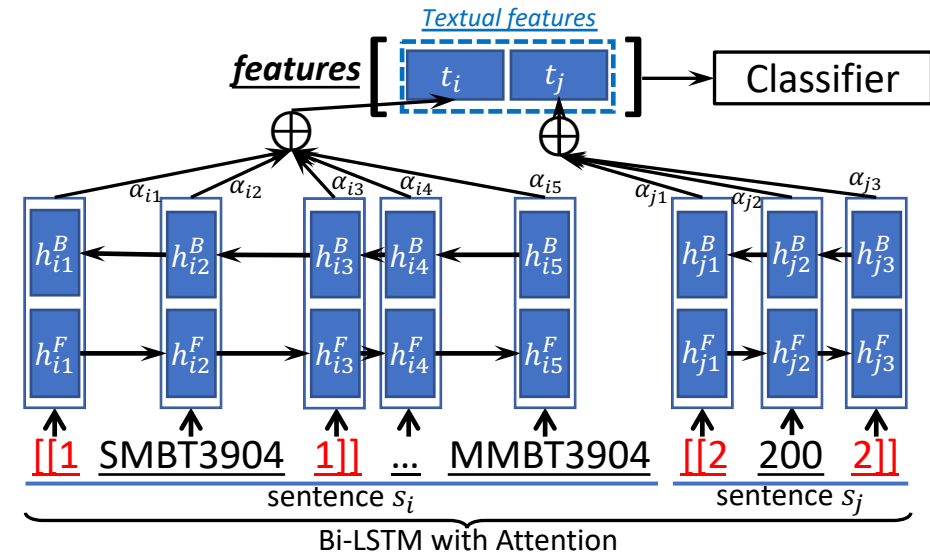
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LSTM excels at relation extraction from text
Xu et al., 2015; Miwa et al., 2016; Zhang et al., 2016



Problem: LSTM networks struggle to capture the multimodal characteristics of richly formatted data.

Augmenting LSTM with Multimodal Features



Transistor Datasheet

Font: Arial; Size: 12; Style: Bold {SMBT3904}...MMBT3904

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Collector current		200	mA
Total power dissipation $T_S \leq 71^\circ\text{C}$ $T_S \leq 115^\circ\text{C}$	P_{tot}	Header: 'Value'; Row: 2; Column: 3 250	
Junction temperature	T_i	150	$^\circ\text{C}$
Storage temperature	T_{stg}	-65 ... 150	

We use the multimodal information stored in the document to extract basic multimodal features:

- Structural features
- Tabular features
- Visual features

Same Font

Aligned

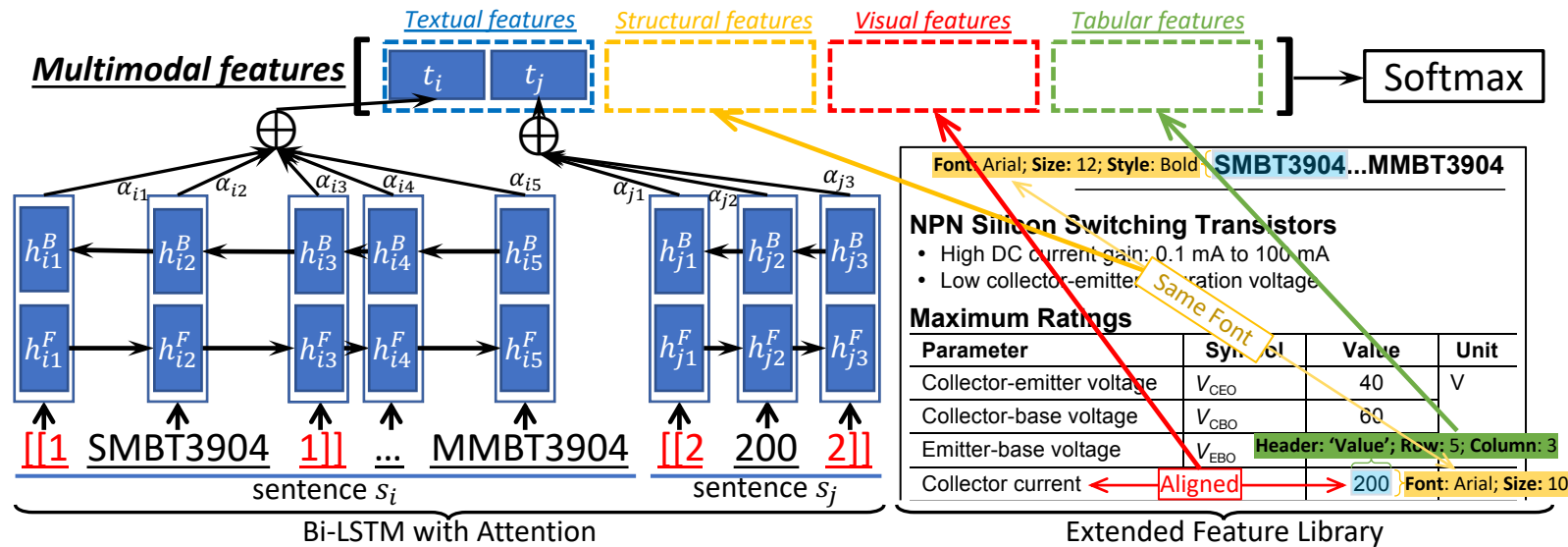
200

Font: Arial; Size: 10

Augmentation with multimodal features captures signals a traditional LSTM would miss.

Fonduer's Multimodal LSTM

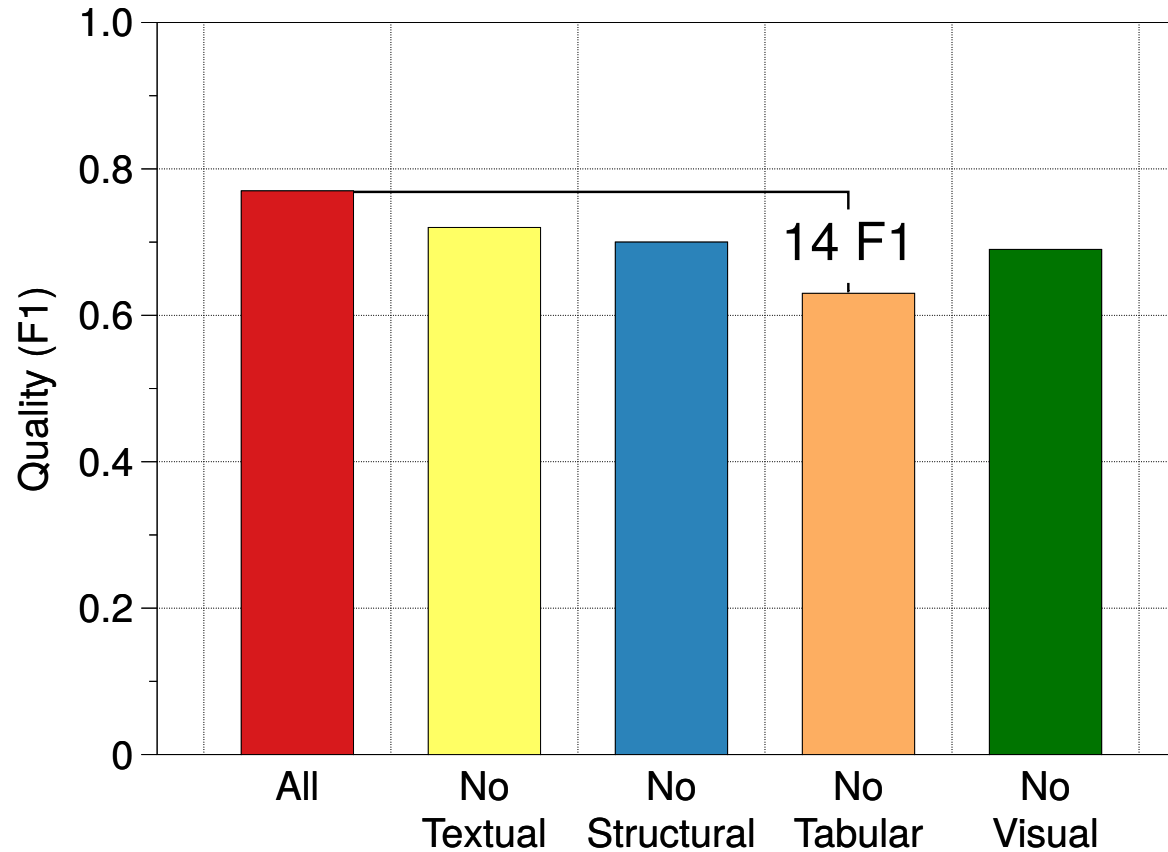
Signals from different modalities can be useful to find the information.



Fonduer: a KBC system that takes advantage of both techniques to reason about all available signals.

The impact of multimodal features

For transistor datasheets...



Multimodal features significantly impact the quality of extraction

Fonduer in the wild

Empirical results & real-world uses

Fonduer vs. Human-curated Knowledge Bases



Fonduer

Same set of documents

Human-created

10 years

1.0x extractions

Machine-created

<6 months

1.59x extractions

Precision **0.89**

How people use Fonduer in industry

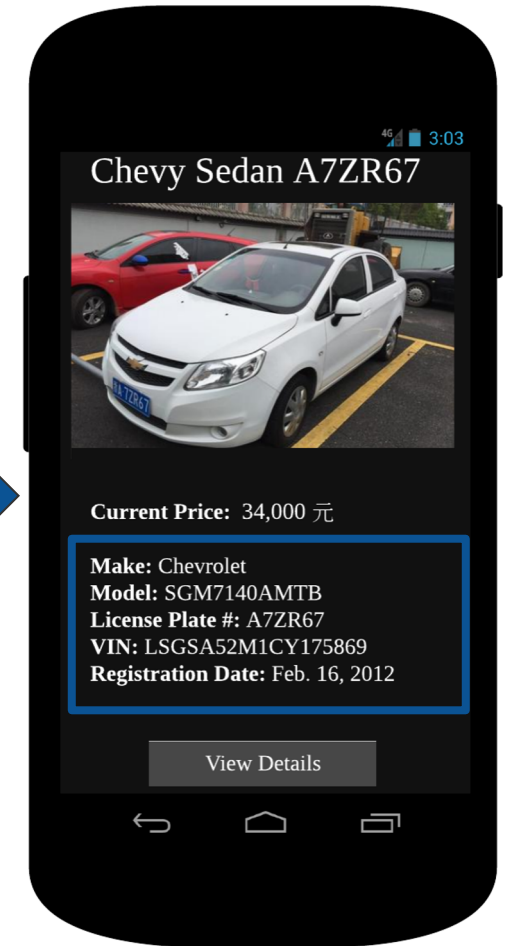
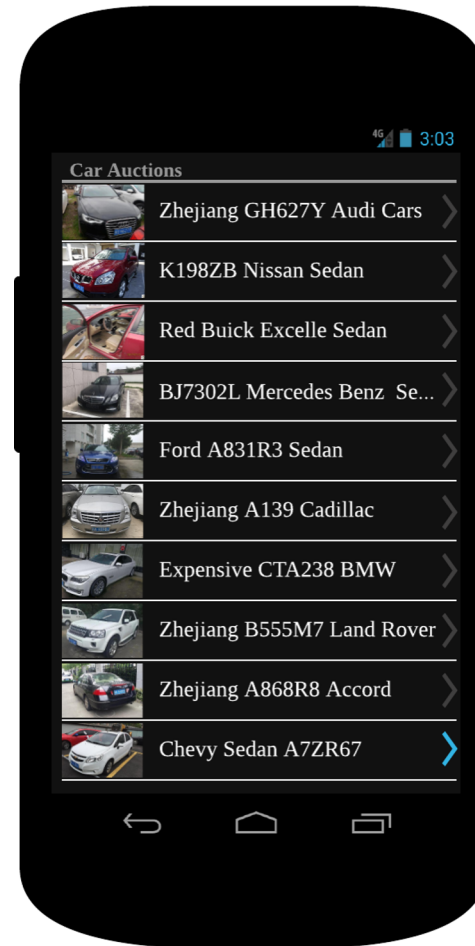
Input: User-customized HTML auction pages → **Output:** Structured knowledge base

Extract key facts (make, model, license, etc.)

Improve auction search quality and UX



Fonduer



Knowledge Base Construction from Richly Formatted Data



- Fonduer helps build high-quality KBC from richly formatted data
- Allows users to leverage multimodal signals
- Augments deep learning model with features from each data modality to achieve high quality
- Fonduer is supporting real world applications

Thank you!
Sen Wu
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