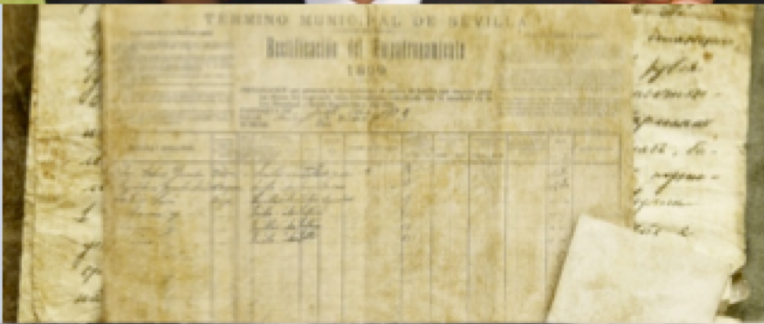


Economical Bimodal Classification of a Massive Heterogeneous Document Collection



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THE CHURCH OF
JESUS CHRIST
OF LATTER-DAY SAINTS


FAMILYSEARCH

Overview

- Timelines (Lead-up)
- Description of the Collections
- Classification Goals for Automation
- Speed-focused System Architectures
- Performance and Outcomes

Timelines (Lead-up)

2015:

FamilySearch was able to auto-*index* 21M born-digital newspapers. Can auto-indexing work with born-paper? How about handwriting??

2016-2017:

FamilySearch & BYU collaborate on technologies to auto-*transcribe* HW.

2017-2018:

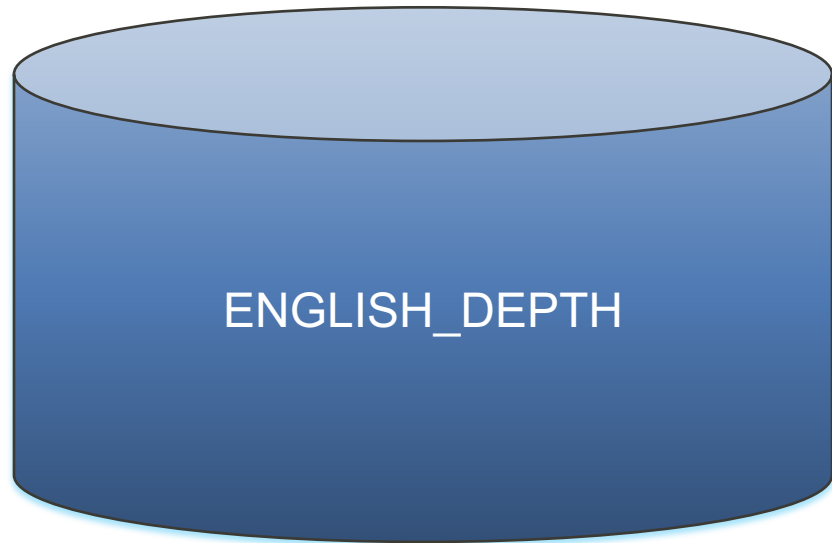
FamilySearch auto-*transcribed* about 33M newspaper stories and over 110M mostly-English handwritten & mixed documents with the goal of auto-*indexing* them.

2019:

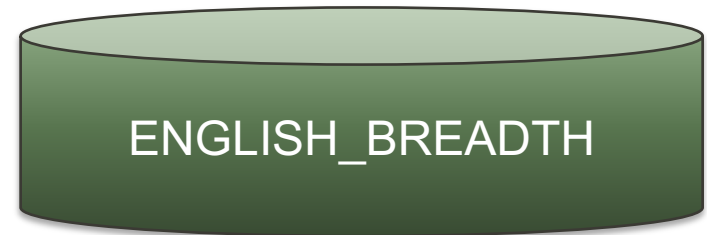
Newspaper going forward. But the massively-heterogeneous collection makes auto-*indexing* complex. Need to group & categorize documents, identify 'gotchas', and subdivide images.

Collections Of Interest

Two different, but related, kinds of corpora:



163K Rolls of Film, every image
[Abt 110M images]
Represents **EVERY instance** of
particular types of US Legal
documents



~1M Rolls of Film, several ims/roll
[Abt 3-4M images]
Represents **EVERY 'English' roll**

Can We Classify After-the-Fact?

If we could **describe each image** of the Breadth/Depth corpora, we could target sub-collections for auto-indexing based on current capabilities & develop the capability for others.

Also, if we could **identify any anomalies**, that might help us do a better job handling them.

But we want to do this **quickly!** **We want to finish in a week or so.** But if we only took 1 sec/document (typical load time of a full image), it'd take
[1.1 x 10⁸ images] x [1 sec/image] = 3.5 CPU years !

Classify: Semantic Categories

130+ Semantic Categories: What is the PURPOSE for the document?

INDICE DECEANALE

ROSE B. CONNORS	AGE	SEX	RESIDENCE	ROSE B. CONNORS	AGE	SEX	RESIDENCE
...

Registration/Civil

Last Will of James M. ...

Probate/Will

STATE OF OHIO
DEPARTMENT OF HEALTH
CERTIFICATE OF DEATH

Vital/Death/Legal

EL SEMANARIO

General/Newspaper

Know all Men by these Presents:

Land/Deed

Family/Pedigree

Family/Pedigree

Classify: Layout Categories

~12 Layout Categories: What is the **STRUCTURE** of the document?

Table/1 Line Per Row

Freeform

(Complex) Form

Multicolumn

Fill in the Blank

Graphical

Classify: Story Count

~12 Story Classification: How many unique 'stories' are in the document?

Story=1n

Story=E&S

Story=1

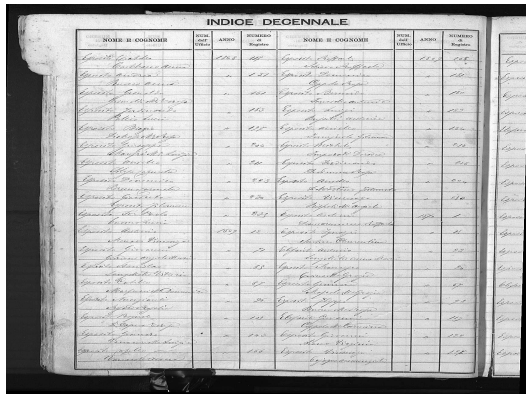
Story=many

Story=2

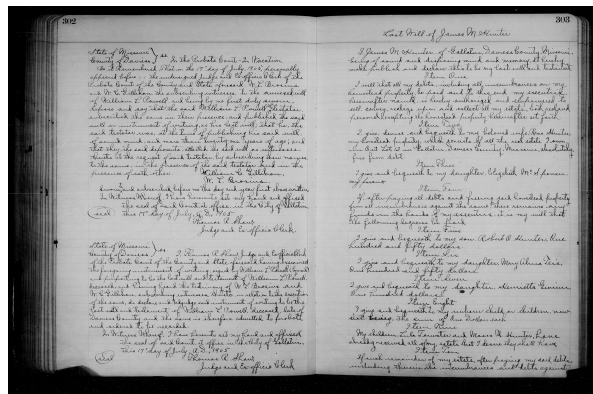
Story=0p

Classify: Language Info

Linguistics: What are the Unicode scripts, language, countries, writing style?



Latin/Italian/MX

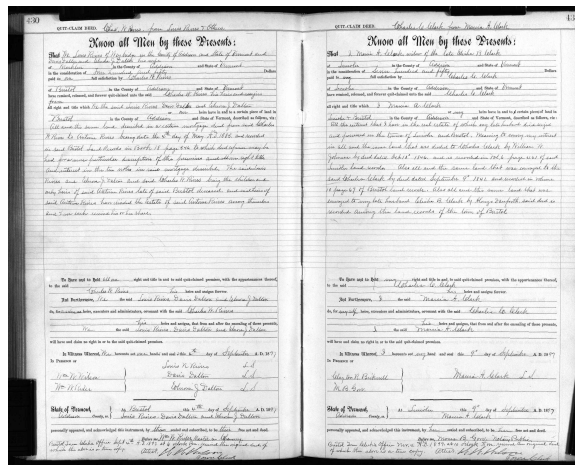


Latin/English/HW

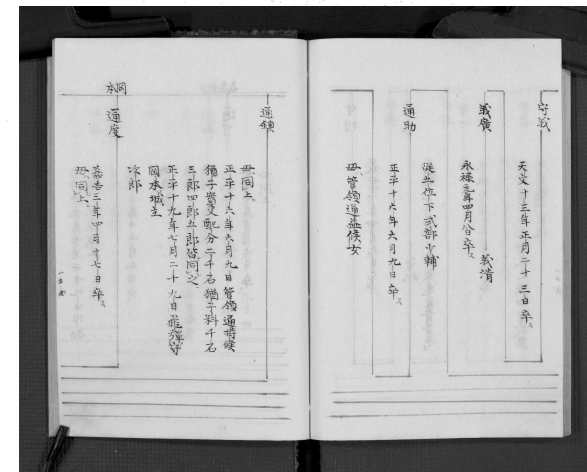
Latin/English/MX



Latin/Spanish/PR



Latin/English/MX



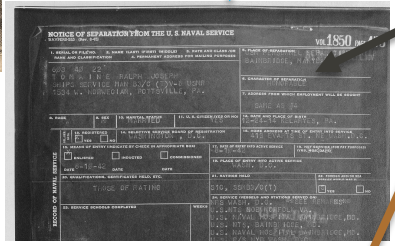
Chinese/Japanese/HP

Anomalies: Binary Properties



SINGLE (with blue arrows pointing left and right)

FOTO (with a blue arrow pointing down to the next image)



REV_VIDEO (with a grey arrow pointing left to the photo)

TWO-D (with an orange arrow pointing down to the next image)

MARGIN (with a pink arrow pointing down to the next image)

DRAW (with a blue arrow pointing down to the next image)

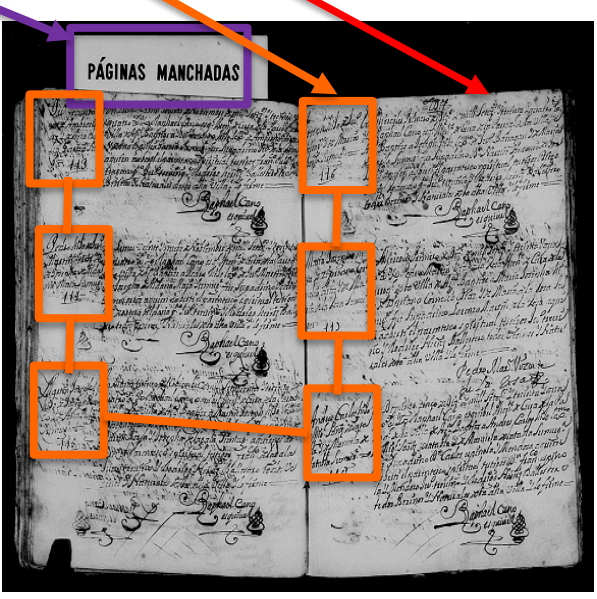
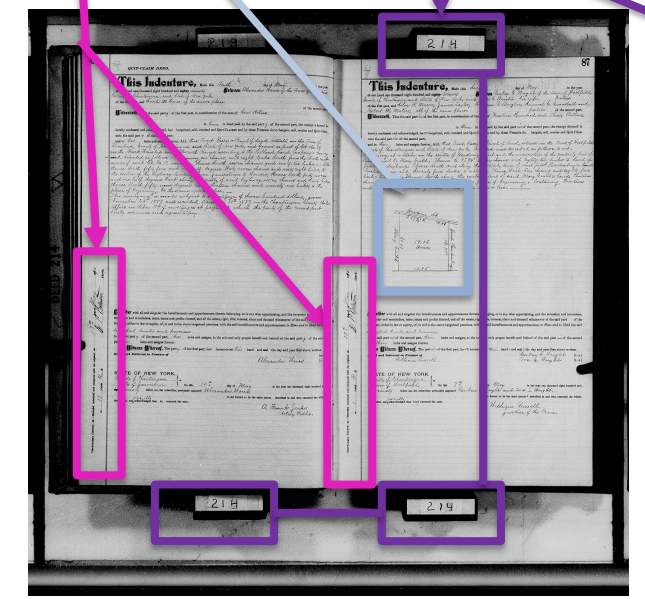
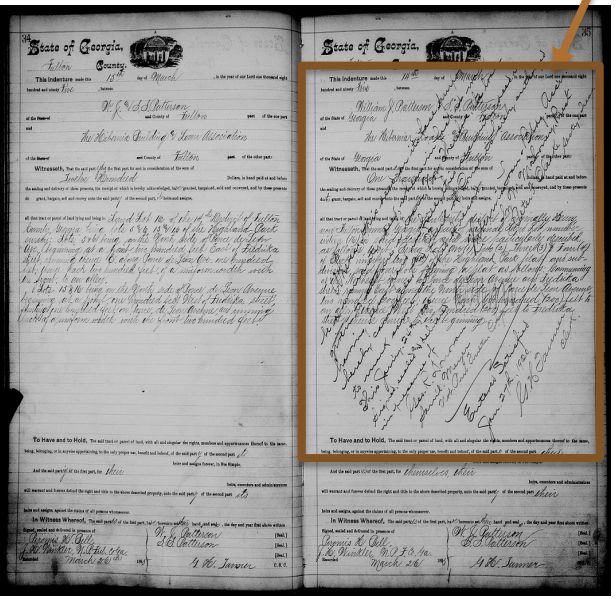
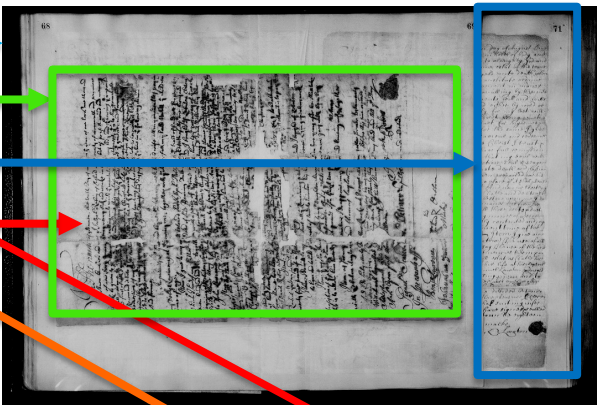
ROTATED (with a green arrow pointing right to the next image)

CRUFT (with a blue arrow pointing right to the next image)

OLD (with a red arrow pointing right to the next image)

LOBE (with an orange arrow pointing right to the next image)

META (with a purple arrow pointing down to the next image)



Speedy Classification?

One Option: Use **thumbnail images** and do image-level classification.

Definite 'Wins':

- FamilySearch automatically stores 200x200 thumbnails of each image.
- Thumbnails for an entire roll of film (1000 images) occupy about the same storage space as 3 images [so, over 99% compression].
- Since these are small, load time and subsequent processing time is short.
- Can see color, periphery, two-up-ness, photos, & line patterns



Table



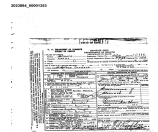
Free



Multi-
column



Paired
Forms



Form



Vertical



RV



Photo

Drawbacks:

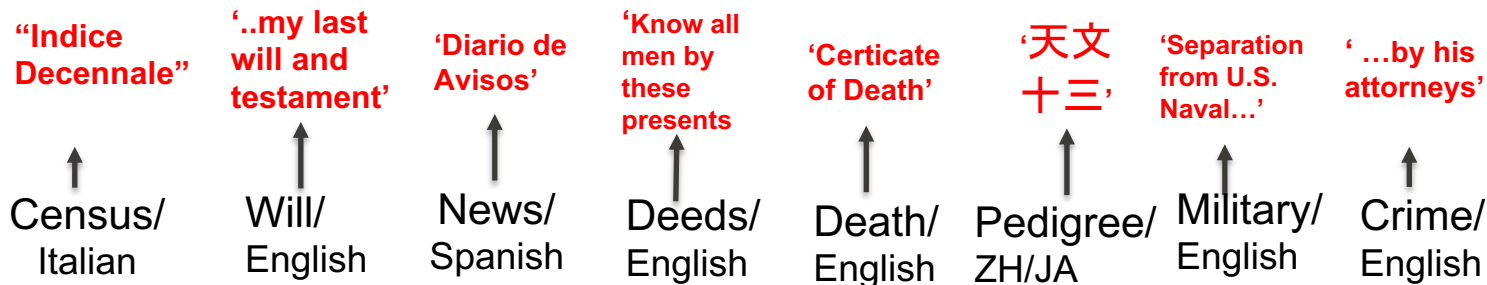
- Their amount of detail is limited, so it's hard to assess the true semantics. Have to guess the semantics based on 'this is a paired form, and that's what deeds look like, so I'll guess it's a deed.'

Speedy Classification?

Another Option: Use **transcripts with bounding boxes** & do text-level classification.

Definite Wins:

- Processing transcript is *orders of magnitude faster* than thumbnails.
- Semantic information is often very clear at the textual level.
- Language, script, country, writing style – should all be straightforward to note.



Serious Drawbacks:

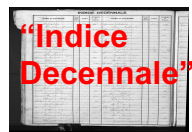
- Color is gone; borders are likely gone; photos are gone. How can one even tell if an image was reverse video if all you have is the transcript? How can you tell if it was complicated form or if it was nicely laid out?
- One needs to have the transcripts already.

Speedy Classification?

BEST Option: Use **BOTH** snapshots AND transcripts+bounding boxes.

Definite Wins:

- Get the best of both worlds: semantics from text, visuals from thumbnail.
- Not much more expensive than JUST thumbnails when using both.
- Can toggle and use text-based or image-based models if that's all one has.



Census/
Italian/
Table



Will/
English/
Free



News/
Spanish
Multicol



Deeds/
English/
PairForm



Death/
English/
Form



Pedigree/
ZH/JA
Vertical



Military/
English/
RV

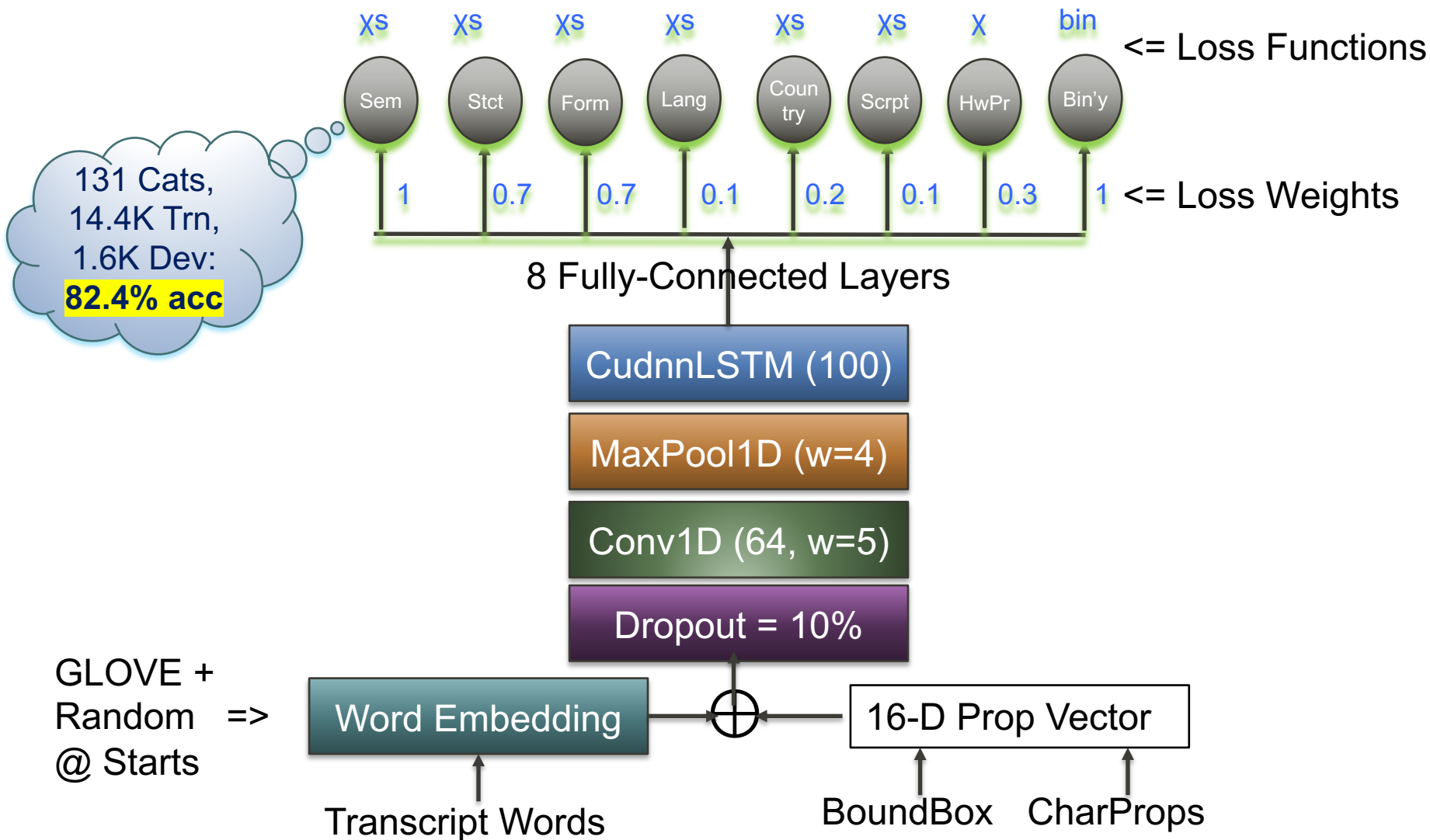


Crime/
English/
Newsclip
w/photo

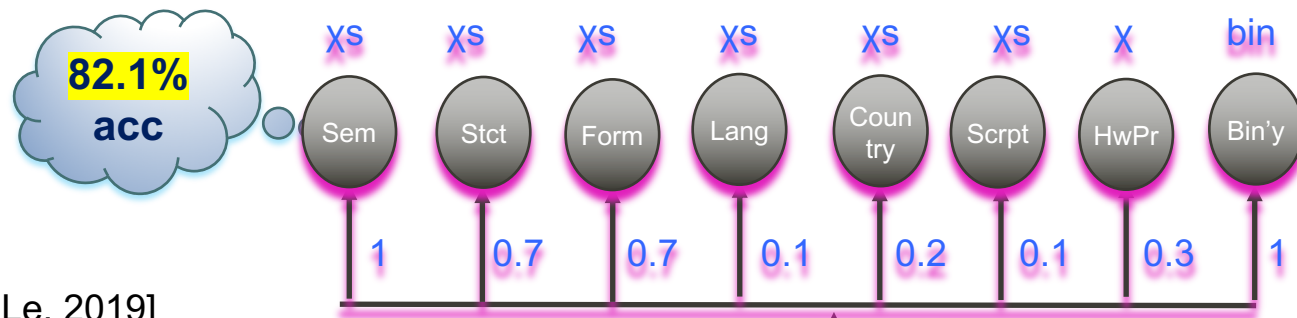
Drawbacks:

- Model management is slightly more complex.

System Architecture: Text Input



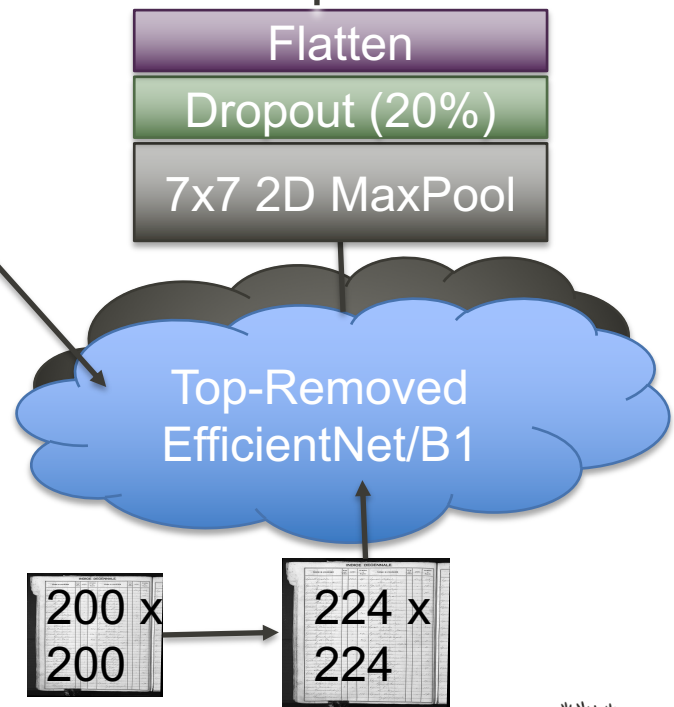
System Design: Image Input



EfficientNet [M. Tan, Q. Le, 2019]

Net	#Param	#Flops	xVersus
B0	5.3M	0.39B	9% (ResNet50)
B1	7.8M	0.70B	12% (Incpt'nV3)
B2	9.2M	1.0 B	7.6% (Incpt'nV4)
B3	12M	1.8 B	5.6% (ResNxt50)
B4	19M	4.2 B	18%(AmoebaNtA)
B5	30M	9.9 B	24%(AmoebaNtC)
B6	43M	19 B	
B7	66M	37 B	

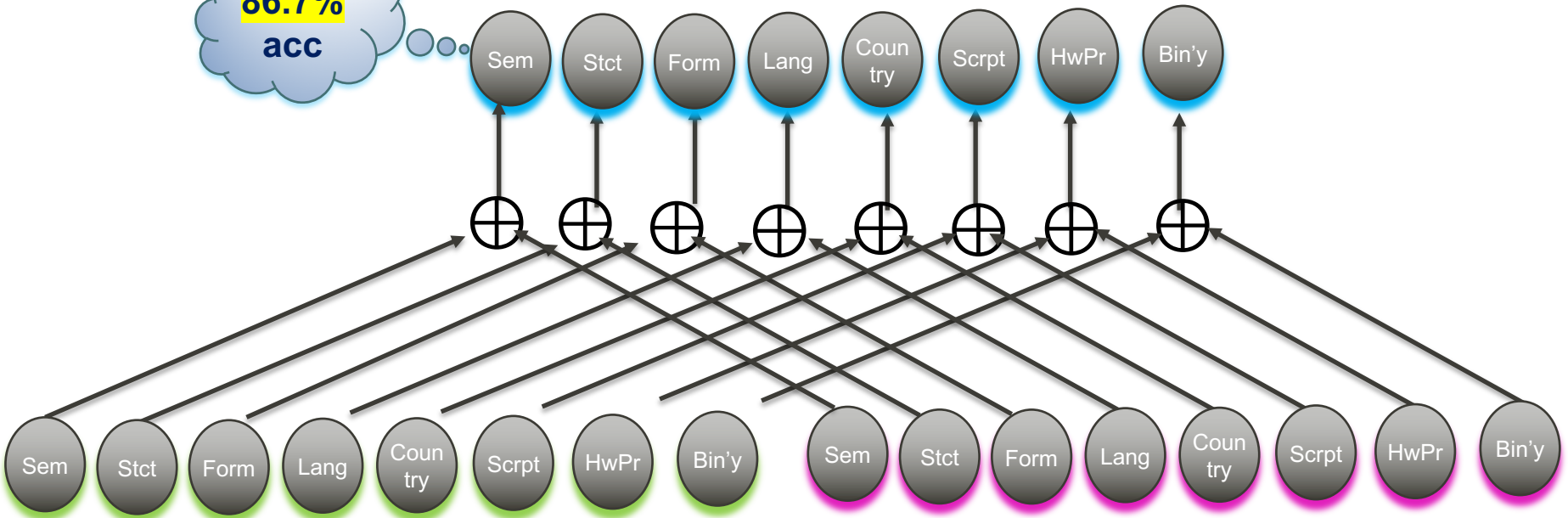
8 Fully-Connected Layers



Results reported by Tan&Le.

System Design: Fused Input

86.7%
acc



For fully-connected weights at start, assume near-50% weights for class C from text(or image) going to class C in final, and near-zero weights for all other connections.

Outcomes: Timings

115,973,482 Images

Ran TWO trials. First was TEXT ONLY, second was FULL.

TextOnly:

Ran on one box (Dual-Gpu System).

Three jobs/Gpu (but lock around Gpu process)

Took 3.5 days.

FullSystem:

Re-Ran on 3 diff't machines, with variable number of Gpus.

But would have taken ~20 days on system of 'TextOnly' (with bulk of the additional cost going to thumbnail processing).

Outcomes: Results

115,973,482 Images

Semantics	%
Deeds	52.6
Land Index	11.6
Gen.Legal	8.3
Gen.Probate	5.6
Will	4.0
Inventory	3.4
Recpt/Check	1.1

Layouts	%
Freeform	68.1
Fill-in	18.2
Table/1line	10.4
Form	1.7

Recording	%
Handwrit'n	59.1
Mixed	22.0
PrintOnly	18.3
Blank	0.7

Anomalies	%
One-ups	52.4
Old (<1800)	3.7
HasMeta	2.0
HasLobes	1.5
ReverseVid	0.6
BleedThru	0.5

#Stories	%
Exactly 1	35.0
EndOrStrt	19.3
>1, but <2	9.3
End&Start	8.4
1-∞ Index	7.7
Exactly 2	7.2
Many	7.0

Summary

- Identified deep neural networks to mine text and image content, with sparse network combiner
- 86.7% acc on 131 category determination, plus generates multiple other kinds of classifications simultaneously
- Demonstrated result on large collection of >110 images

QUESTIONS?