

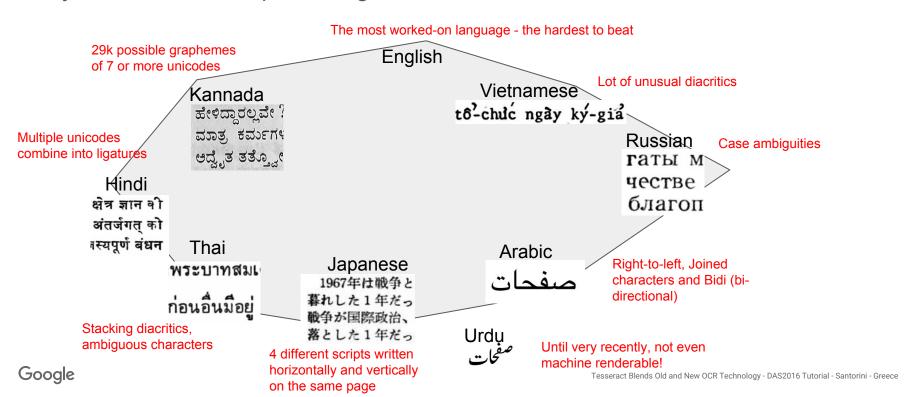
# 7. Building a Multilingual OCR Engine

Training LSTM networks on 100 languages and test results

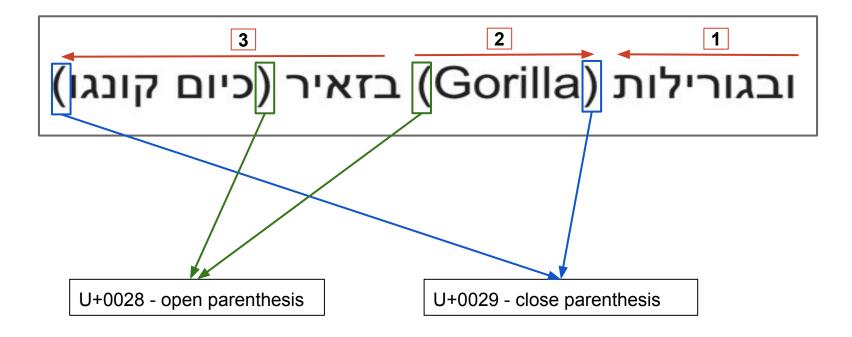
Ray Smith, Google Inc.

#### Internationalization: The Convex Hull of Languages

If you want to develop multilingual OCR, work on these first...



#### **Bidirectional issues**



## What's a "character" in Devanagari?

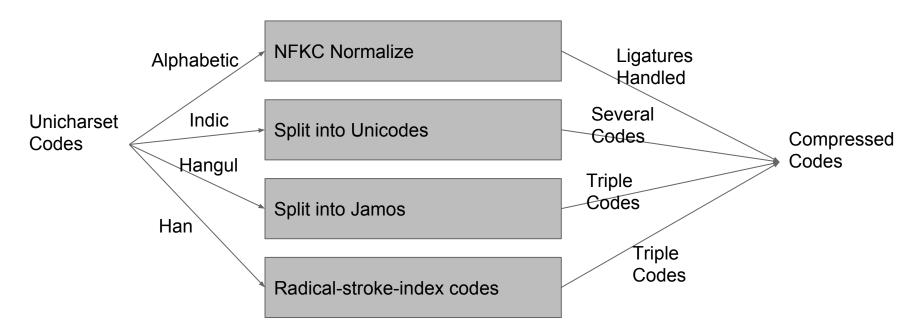
Result	Unicode	Transliteration
र	0930	ra
र्	0930 094d	r
र्द	0930 094d 0926	rda
दं	0930 094d 0926 094d	rd
र्द्ध	0930 094d 0926 094d 0935	rdva
र्द्धि	0930 094d 0926 094d 0935 093f	rdvi
र्द्धिक	0930 094d 0926 094d 0935 093f 0915	rdvika

#### What's a "character" in Kannada?

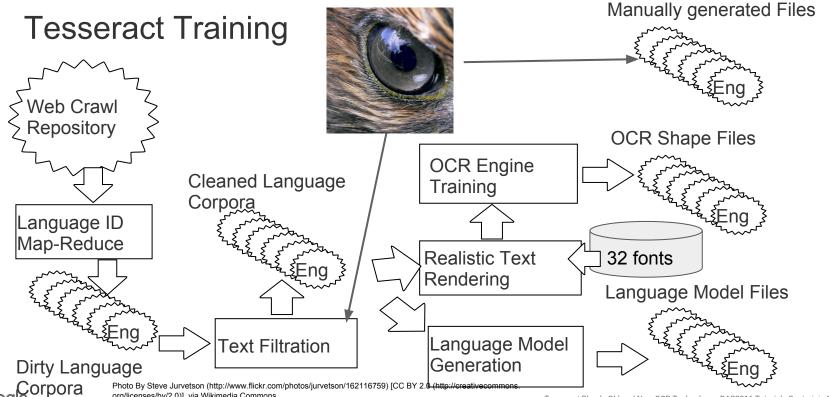
Result	Unicode	Transliteration
ರ	0cb0	ra
ದ	0ca6	da
ವ	0cb5	va
ਰਾ	0cb0 0ccd	r
ರ್ದ	0cb0 0ccd 0ca6	rda
ರ್ದ್	0cb0 0ccd 0ca6 0ccd	rd
ದರ್ವ ದ್ವ್	0cb0 0ccd 0ca6 0ccd 0cb5	rdva
ದರ್ವ್ ರ್ದ್ವ್	0cb0 0ccd 0ca6 0ccd 0cb5 0ccd	rdv
ದರ್ವ್ಕ್ ದ್ವ್ಕ್	0cb0 0ccd 0ca6 0ccd 0cb5 0ccd 0c95	rdvka
ದಿರ್ವ್ಯ ದ್ವ್ಯೀ	0cb0 0ccd 0ca6 0ccd 0cb5 0ccd 0c95 0cbf	rdvki

#### Universal Character/Grapheme Encoding/Compression

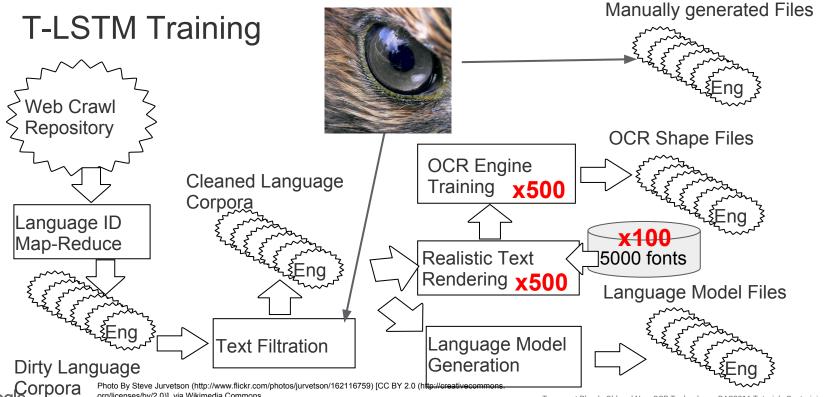
Extension to Tesseract's UNICHARSET to make the output Softmax smaller



### Training International OCR Engines



### Training International OCR Engines



#### Training

Synthetic Training data:
With bounding boxes
Instead of CTC

hearing 6 REFERENCES \$1.99 := board field & Spring GmbH (Publisher) = USEFUL

#### Food lyrics comment?

ಕಾರ್ಬೊನೇಷನ್ ವರ್ಣಾಂಶವು ಮೇಲೂ ಹ್ರಸ್ವರೂಪವೆಂದು ಬಿಲ್ಫಾಸ್ಟ್ನ ಪ್ರಾರಂಭಿಕ ಮದ್ರಾಸ್ಗೆ ಪೊಡ್ನಿಕ್ ಕಟ್ಟುತ್ಲಿ

เชียงใหม่ สินธุ ผู้ระบบง่ายๆ บ็อบ <u>ศรีอยุธยา ชานี่ทั้ง</u> อ้างอิงคือ๒๕๔๘ 2

- About 500k lines per language
- Random book-like degradation
- (Almost) Same network specification for each language:
- Convergence in 3-5 days or more

[G2,0C2,2FT16P3,3LQ1,64L1,128RtL1,128LS1,256] [G2,0C2,2FT16P3,3LQ1,64L1,128RtL1,128LS1,512]

#### **Testing**

#### Testset from Google Books:

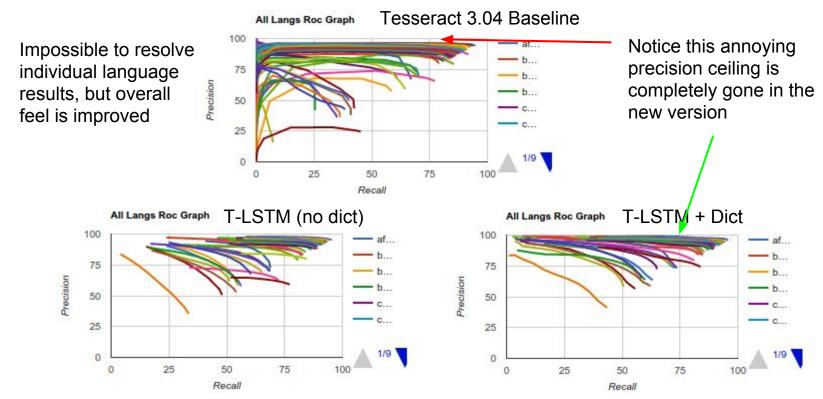
- Single Lines cut from older books.
- Hand typed. Accuracy far from perfect.
- 1000 lines \* ~50 languages.

Caveat: Does not allow Tesseract to adapt to a whole page. (T-LSTM doesn't adapt)

#### Example:

carried out by a single artist. He observes that the greater part is of a marble

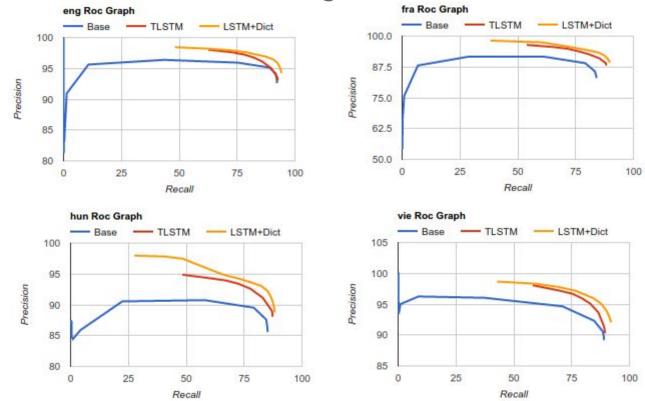
#### Overall effect on 51 Languages



## Accuracy Results: Selection of Latin Languages

	Truth								
Lang	Words	Char Er	ror Rates		%Change	Word Error Rates			%Change
		Base	T-LSTM	LSTM+Dict		Base	T-LSTM	LSTM+Dict	
Czech	19319	3.08	1.86	1.81	-41.23	12.69	8.26	7.58	-40.27
English	21543	2.51	1.95	1.76	-29.88	7.58	7.18	5.77	-23.88
French	20746	5.98	3.1	2.98	-50.17	16.63	11.82	10.47	-37.04
Hungarian	17977	3.75	2.95	2.86	-23.73	14.61	12.26	11.45	-21.63
Indonesian	16616	2.77	1.96	2.26	-18.41	9.45	8.64	7.51	-20.53
Dutch	18878	6.23	6.65	5.97	-4.17	16.38	17.55	15.75	-3.85
Norwegian	18129	2.19	1.74	1.81	-17.35	7.73	5.87	5.7	-26.26
Portuguese	17726	2.91	1.9	1.87	-35.74	10.09	7.46	6.75	-33.10
Spanish	20333	4.28	2.5	2.31	-46.03	11.63	8.55	7.45	-35.94
oo⁄gietnamese	22472	4.04	3.55	2.61	-35.40	10.89	t Blends 1 On 05	DCR Technology - DAS 701 $93$	-27.18

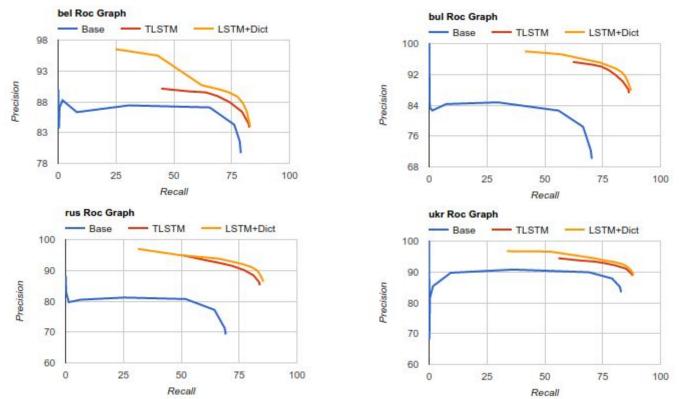
#### Accuracy Results: Latin Langs ROC curves



#### Accuracy Results: Cyrillic Languages

Lang	Truth Words	Char E	rror Rate	S	%Change	Word Error Rates			%Change
		Base T-LSTM LSTM+Dict				Base	T-LSTM	LSTM+Dict	
Belarusian	11697	7.72	4.25	4.60	-40.41	20.63	16.66	16.38	-20.60
Bulgarian	18457	19.68	3.78	3.85	-80.44	29.88	13.18	12.4	-58.50
Macedonian	17117	11.22	3.18	3.08	-72.55	18.92	11.69	10.61	-43.92
Russian	14993	19.06	4	4.10	-78.49	30.83	15.34	14.04	-54.46
Ukrainian	17123	6.63	2.77	3.27	-50.68	16.78	11.6	11.34	-32.42

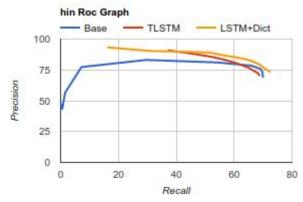
## Accuracy Results (Cyrillic Langs)

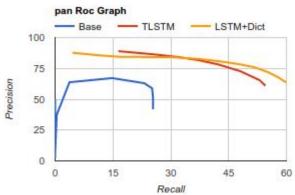


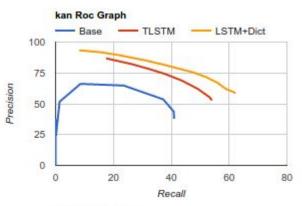
# Accuracy Results (Indic Languages)

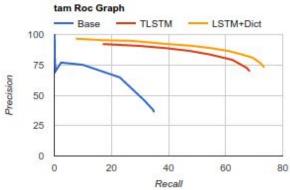
Lang	Truth Words	Char Error Rates			%Change	Word Error Rates			%Change
							T-		
		Base	T-LSTM	LSTM+Dict		Base	LSTM	LSTM+Dict	
Bengali	16865	19.59	22.33	19.25	-1.74	42.79	42.23	39.13	-8.55
Gujarati	19874	45.7	22.38	18.35	-59.85	56.46	49.46	43.62	-22.74
Hindi	27539	14.05	13.56	11.68	-16.87	30.92	30.19	26.92	-12.94
Kannada	13673	31.55	13.57	12.12	-61.58	63.2	47.11	40.71	-35.59
Marathi	21486	29.04	12.07	9.18	-68.39	47.19	32.79	25.92	-45.07
Nepalese	20606	29.56	18.38	15.63	-47.12	50.19	42.39	36.59	-27.10
Panjabi	27651	46.23	19.89	16.1	-65.17	54.89	40.18	37.15	-32.32
Tamil	10033	26.1	10.09	9.3	-64.37	63.58	30.65	26.78	-57.88
Telugu	14133	31.47	31.51	27.8	-11.66	63.63	63.79	59.36	-6.71

## Accuracy Results (Indic Langs)





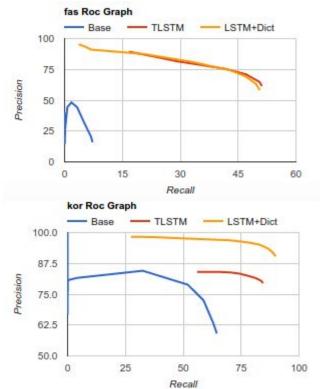


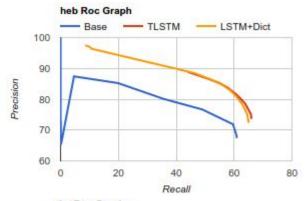


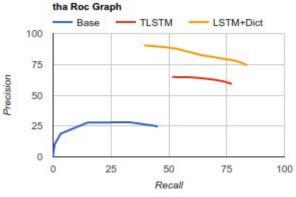
### Accuracy Results (Other Languages)

Lang	Truth Words	Char E	rror Rate	S	%Change	Word Error Rates			%Change
		Base	T-LSTM	LSTM+Dict		Base	T-LSTM	LSTM+Dict	
Hebrew	21919	12.29	8.08	8.36	-31.98	34.28	28.81	29.91	-12.75
Thai	32173	60.52	12.32	7.73	-87.23	96.91	38.46	22.70	-76.58
Yiddish	21674	20.76	10.04	10.62	-48.84	56.77	34.32	36.53	-35.65
Farsi	23691	74.87	16.53	16.84	-77.51	65.44	40.43	42.80	-34.60
Chinese(S)	31045	8.25	7.25	6.42	-22.18	10.82	11.41	9.32	-13.86
Chinese(T)	28700	13.22	12.83	10.35	-21.71	18.56	20.28	15.78	-14.98
Japanese	29574	18.65	16.97	11.53	-38.18	31.66	35.49	19.94	-37.02
Korean	25687	31.19	9.62	6.67	-78.61	40.09	18.69	9.82	-75.51

#### Accuracy Results (Other Langs)







#### Conclusion

- Tesseract is surprisingly general
- T-LSTM is much better almost across the board
  - Some Language model integration issues remain
- Some languages remain difficult, but
- Neural networks are taking over!

#### Thanks for Listening!

# Questions?