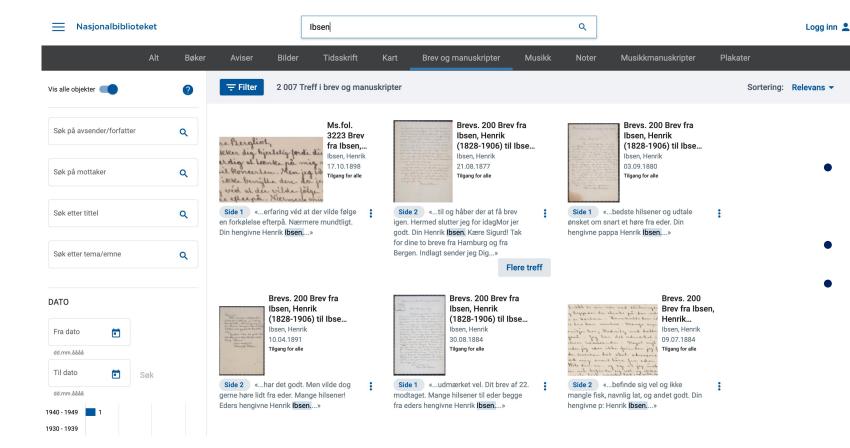
# A Comprehensive Comparison of Open-Source Libraries for Handwritten Text Recognition in Norwegian

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DAS 2022 – La Rochelle



# Searching the manuscripts at Nasjonalbiblioteket



- Searching in meta-data and full text
- Provide faceted search
- Index persons, places, time, etc

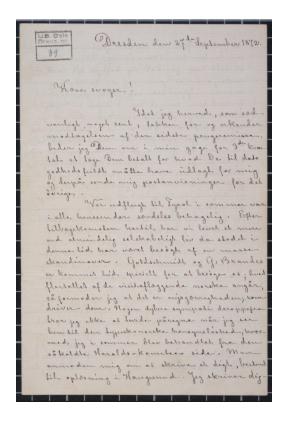
## HTR at the Nasjonalbiblioteket

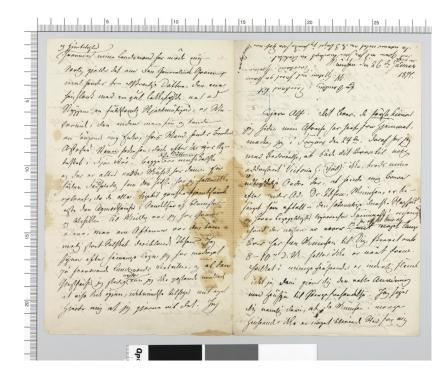
#### Objectives:

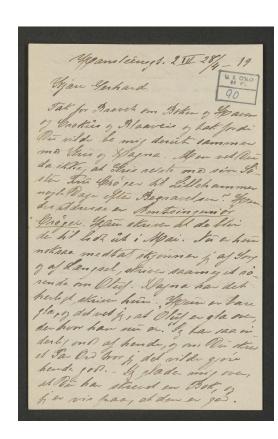
- Include handwriting recognition in the standard digitization process
- Use open-source software for document processing
- Produce resources for HTR in Norwegian
- Develop and formalize best practices for HTR



#### The NorHand Dataset







Letters from Henrik Ibsen (1872), Camilla Collett (1877) and Harriet Backer (1919).



#### The NorHand Dataset

Writer	Lifespan	Rand	dom s	plit	Wri	$_{ m lit}$	
		train	val	test	train	val	test
Backer, Harriet	1845-1932	58	9	10	58	9	0
Bonnevie, Kristine	1872-1948	43	5	5	43	5	0
Broch, Lagertha	1864-1952	43			43		
Collett, Camilla	1813-1895	68	10	10	68	10	0
Garborg, Hulda	1862-1934	166	30	16	166	30	0
Hertzberg, Ebbe	1847-1912	48	6	6	48	6	0
Ibsen, Henrik	1828-1906	42	4	5	42	4	0
Kielland, Kitty	1843-1914	34	5	5	0	0	44
Munch, Edvard	1863-1944	33	5	5	0	0	43
Nielsen, Petronelle	1797-1886	58			58		
Thiis, Jens	1870-1942	41	4	4	41	4	0
Undset, Sigrid	1882-1949	40	5	5	0	0	50
Total		674	83	71	567	68	137

	Pages	Lines	Words	Chars
Train set	674	19,653	139,205	637,689
Validation set	83	$2,\!286$	13,916	$61,\!560$
Test set	71	1,793	11,801	$52,\!831$
Total	828	23,732	164,922	752,080

- Manual transcription at line level
- Available in Page XML format
- Official splits provided
- Version 1 (more to come)

Download: https://zenodo.org/record/6542056



### Survey of recent open source HTR libraries

- Survey of HTR libraries used in IJDAR, ICDAR, ICFHR, DAS, ICPR papers
- Between 2019 and 2021
- Open source
- Compared to state-of-the-art systems on publicly available databases of handwritten documents in European languages

10 libraries + HTR+ from Transkribus



### Selection of open source HTR libraries

#### Selected

Name	Framework	Last commit	Commits	Contrib.	Last version
Kaldi [1]	Kaldi	18/12/2021	9223	100	-
Kraken [13]	PyTorch	19/12/2021	1486	18	11/2021
PyLaia [24]	PyTorch	08/02/2021	860	4	12/2020
$\overline{\text{HTR-Flor}++[20]}$	TensorFlow 2	8/12/2021	280	4	10/2020
PyTorchOCR [4]	PyTorch	10/09/2021	24	1	-
VerticalAttentionOCR [5]	PyTorch	3/12/2021	21	1	-
Convolve, Attend & Spell [12]	PyTorch	24/06/2019	20	2	-
HRS[3]	TensorFlow	19/03/2021	20	2	-
ContentDistillation [11]	PyTorch	13/06/2020	3	1	-
Origaminet [28]	PyTorch	13/06/2020	2	2	-
HTR+ [17]	-	-	NA	NA	-

and HTR+

- Number of commits: active development
- Number of contributors: future maintenance
- Date of last commit: recently updated
- Date of last version/package: best practice of software development



## Training of HTR models

- We trained the models from bounding boxes and manual transcriptions
- For each library, 2 setups:
  - Basic model: from the documentation (non-expert)
  - Expert model: with the support of the creators of the libraries
- Vertical lines are ignored
- Training with random split and writer split



# Recognition results (random split)

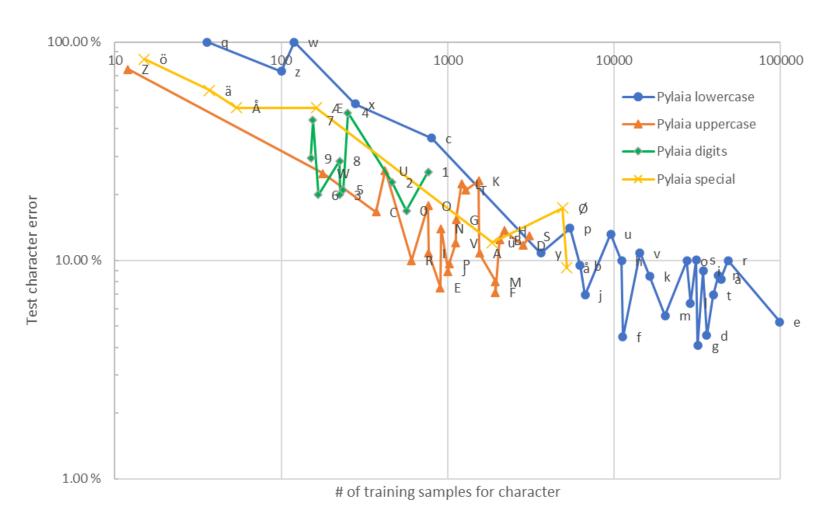
Model	Height	Augm.	Train Val		Test			
			CER	WER	CER	WER	CER	WER
Kaldi basic	40	no	5.30	12.05	11.61	26.19	10.76	24.85
Kaldi expert	40	no	4.71	11.10	10.29	24.17	9.18	22.19
Kraken basic	48	no	51.95	76.52	64.60	89.72	64.44	89.49
Kraken expert	120	yes	0.40	1.31	12.05	30.29	12.20	31.28
PyLaia basic	128	no	1.37	4.45	11.02	28.09	10.87	27.62
PyLaia basic	128	yes	3.08	9.39	10.44	26.50	10.10	26.30
PyLaia expert	64	yes	3.73	10.66	11.70	28.90	12.75	31.12
PyLaia expert	128	yes	1.68	5.30	9.15	24.28	8.86	23.79
HTR-Flor++ basic	128	yes	-	-	-	-	11.49	31.59
HTR-Flor++ expert-a	128	yes	_	_	_	_	56.10	82.21
HTR-Flor++ expert-b	128	yes	-	_	_	_	12.62	32.33
HTR-Flor++ expert-c	128	yes	-	-	-	-	11.04	29.70
HTR+ basic	N/A	N/A	2.98	-	7.17	-	9.14	21.81
HTR+ expert	N/A	N/A	2.58	-	6.34	-	8.31	20.30

Help of an expert is usefull

Data augmentation improves the model



# Detailed CER analysis



Pylaia Expert model

No language model

Strong correlation between CER and number of training samples

#### Most common confusion

Char	# Confusions	Relative confusion	Conf. 1		(	Conf. 2		Conf. 3	Others
a	271	7.38 %	О	2.9 %	е	1.93 %	æ	0.79 %	1.77 %
b	42	8.08 %	l	2.9 %	t	1.54~%	h	1.35 %	2.31 %
e	207	2.60 %	a	0.5~%	О	0.39 %	i	0.29 %	1.46 %
h	86	8.13 %	S	2.5~%	t	1.13 %	k	0.85 %	3.69 %
$\overline{\mathrm{m}}$	74	4.49 %	n	2.61 %	V	0.61 %	i	0.24~%	1.03 %
$\overline{n}$	189	5.59 %	r	1.72 %	m	1.18 %	V	0.68 %	2.01 %
0	162	7.98 %	a	3.20 %	e	1.87 %	Ø	1.04 %	1.87 %
$\overline{r}$	198	5.18 %	S	0.89 %	n	0.89 %	V	0.55~%	2.85 %
S	188	7.25 %	r	1.74 %	h	1.04 %	e	0.81 %	3.66 %
$\overline{\mathrm{F}}$	5	5.21 %	Τ	2.1 %	f	1.04 %	d	1.04 %	1.04 %
$\overline{}$	13	20.00 %	t	9.2 %	1	3.08 %	d	3.08 %	4.62 %
æ	34	7.93 %	e	2.3~%	a	2.10 %	d	0.93 %	2.56 %
Ø	56	14.74 %	О	6.1 %	å	2.37 %	e	1.58 %	4.74 %
$ {a}$	21	11.60 %	Ø	4.4~%	a	3.32 %	u	1.11 %	2.76 %

#### Pylaia expert model



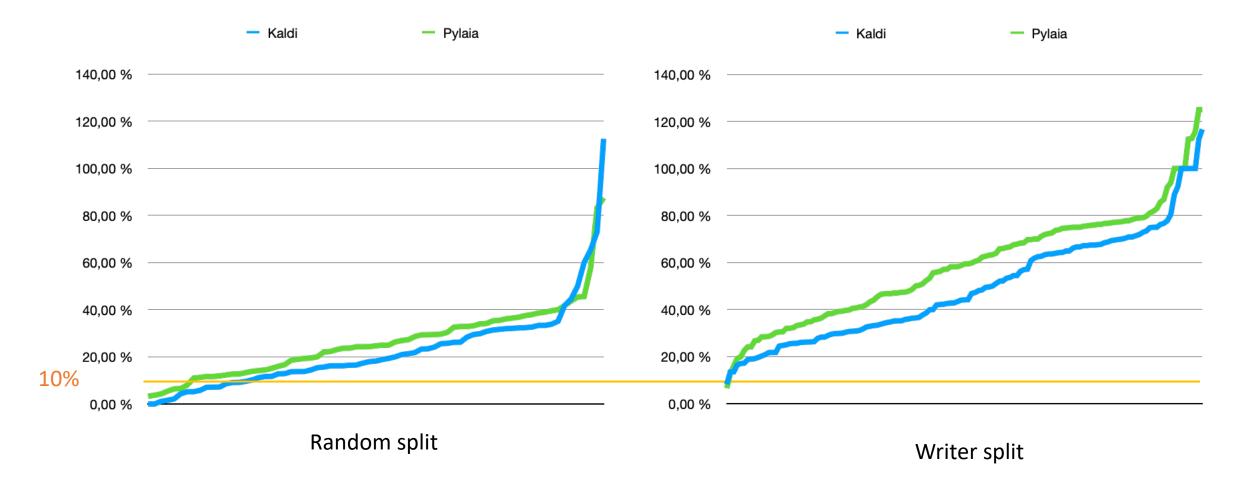
### Recognition results with unseen writers split

Model	Height	Augm.	$\operatorname{Tr}$	ain	n Val		Test	
			CER	WER	CER	WER	CER	WER
Kaldi basic	40	no	4.90	11.34	12.57	28.10	24.24	44.49
Kaldi expert	40	no	4.37	10.48	11.03	25.79	21.79	42.13
PyLaia basic	128	yes	2.70	8.25	10.64	27.58	24.36	49.42
PyLaia expert	128	yes	1.64	5.40	9.53	25.90	22.74	47.95

- Training the best models with the writer split
- Lack of generalization, not enough different writers



#### Distribution of WER at document level





#### Conclusions

- New challenging dataset for HTR
- Comparison of open source HTR libraries with software criteria and CER/WER
  - need to promote best practices in software development for HTR libraries
- Need to go beyond CER/WER analysis
- No Transformer: did not meet the criterion, but to be updated



# Tusen takk!

