

Importance of Textlines in Historical Document Classification

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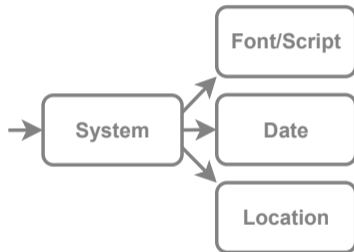
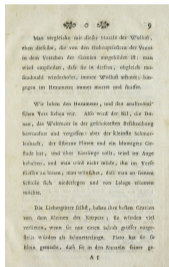


Page-level label(s) without any further details

Font & script classification
Up to N possible labels per page

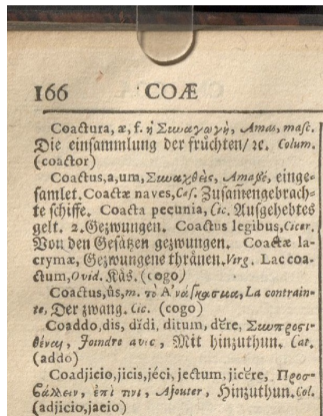
Localization
One label per page

Dating
(*not-before*; *not-after*) intervals



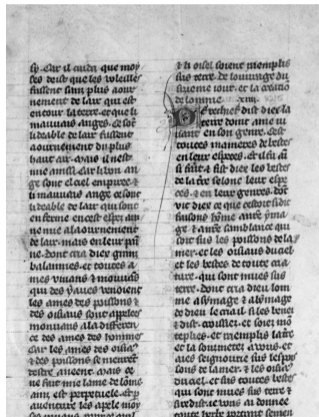
Font/script

Fraktur, greek, italic, antiqua



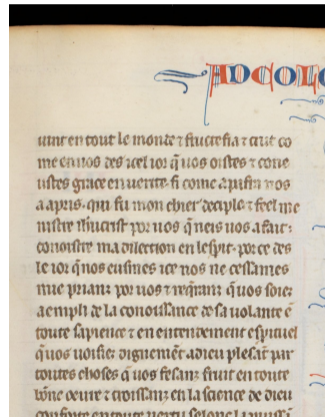
Localization

Paris



Dating

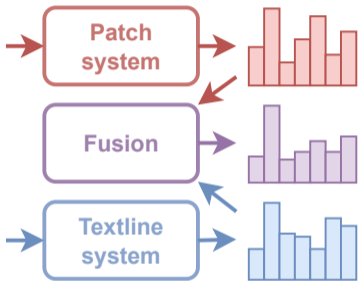
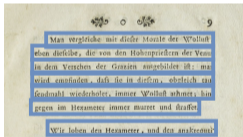
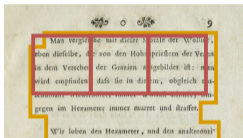
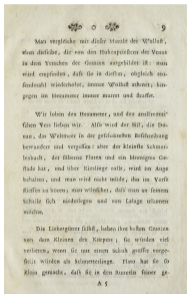
1260-1269

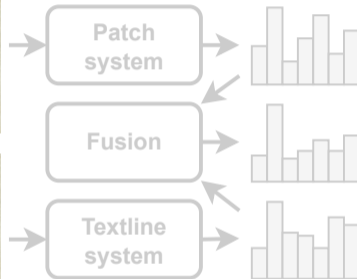
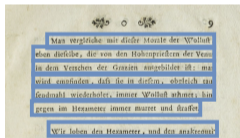
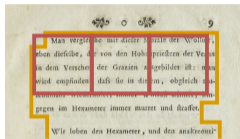
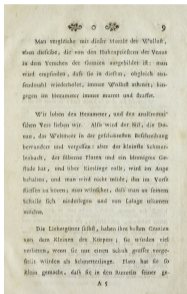


- custom training-validation dataset splits
- the aim to create class-balanced validation set
- publicly available at pero.fit.vutbr.cz/hdc_dataset ↪

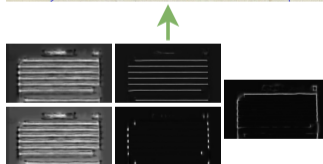
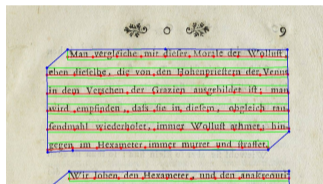
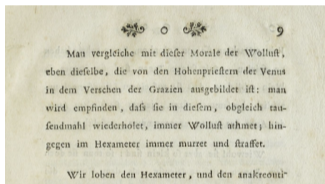
Task	Training	Validation	Testing
Font	35 382	239	5 506
Script	7 594	419	1 256
Location	5 397	65	325
Date	10 294	1 000	2 516



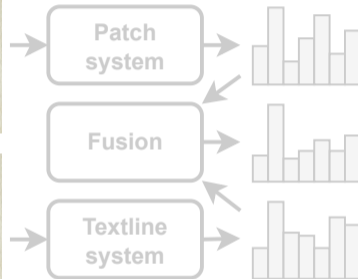
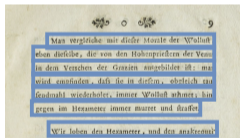
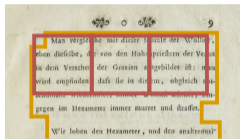
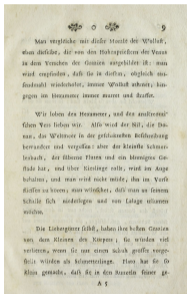


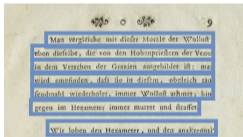
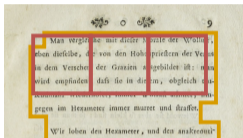
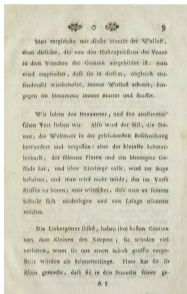


- ParseNet (U-net-based) neural network¹
- Text lines and regions detection
- Trained on PERO Layout dataset

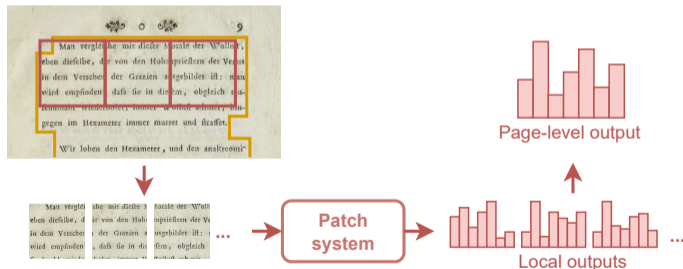


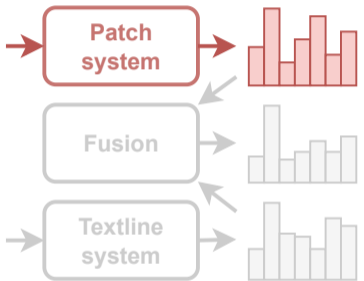
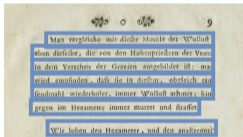
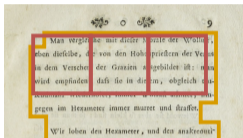
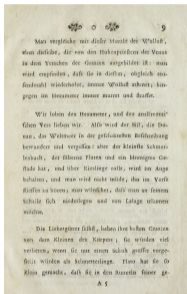
¹Kodym, Oldřich, and Michal Hradiš. "Page layout analysis system for unconstrained historic documents."

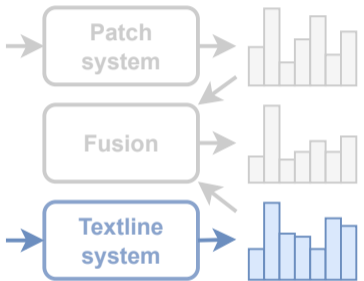
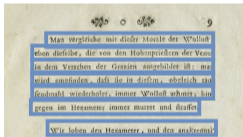
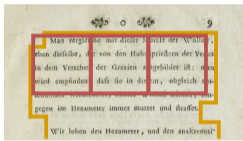
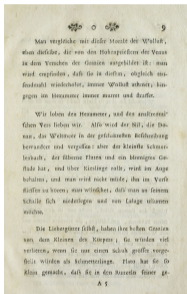




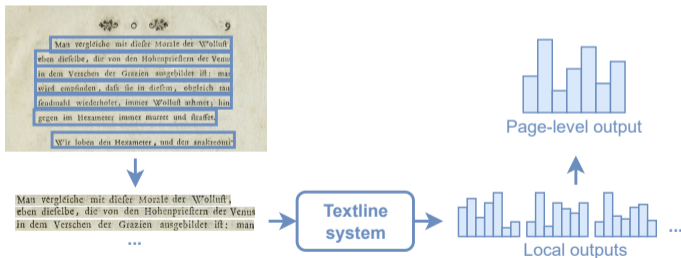
- ResNeXt-50-based neural network
- Training on **four scales** of non-overlapping square patches with **three training strategies**
- The page-level output is calculated as **the mean of the local outputs**

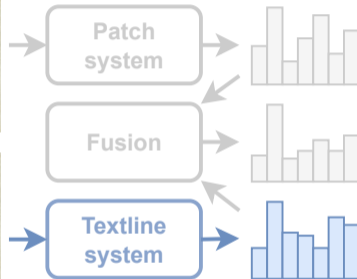
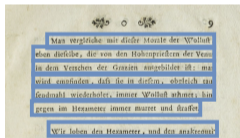
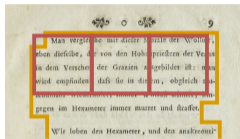
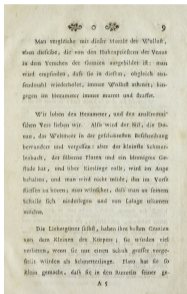


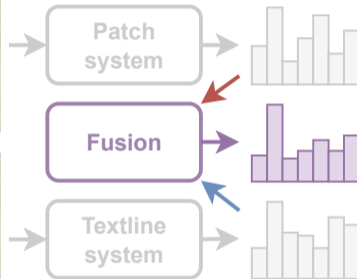
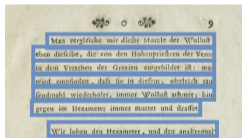
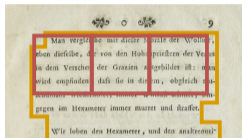
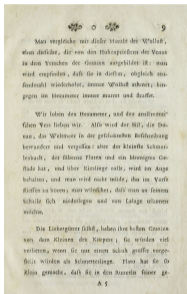




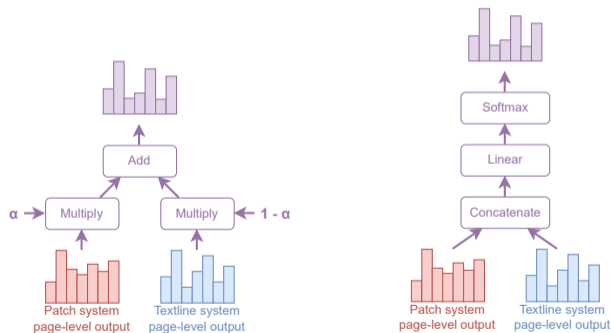
- VGG-based neural network + global average pooling
- Arbitrarily long height-normalized **textline image** as an input
- The page-level output is calculated as the **mean (classification tasks)** or **median (dating task)** of the local outputs

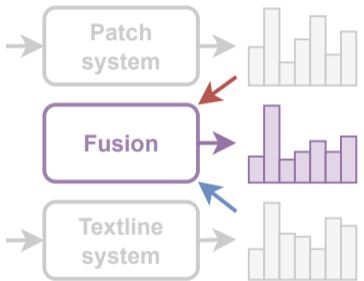
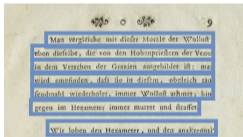
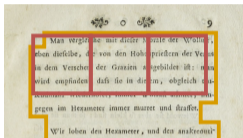
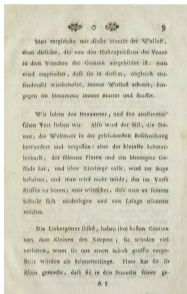


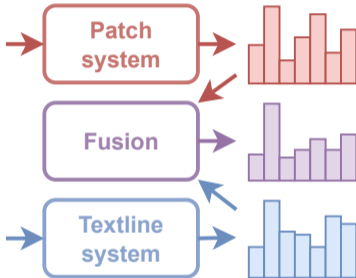
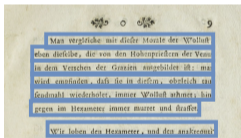
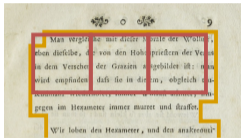
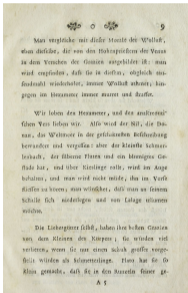




- Linear (left image) and log-linear (right image) fusions
- Page-level outputs from the patch and the textline systems as the input
- Optimized on the validation sets using 10-fold cross validation





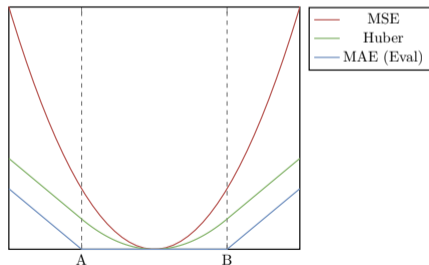


- Up to N possible fonts/scripts per page
 - e.g. $\mathcal{T} \in \{\textit{fraktur}, \textit{greek}, \textit{italic}, \textit{antiqua}\}$
- Page-level annotations **without location information**
- L_{hard} selects the most probable page-level label based on network output
- L_{soft} considers all page-level labels weighted by the network output probability

$$L_{hard} = \min_{i \in \mathcal{T}} [-\log(f(x)_i)] \quad (1)$$

$$L_{soft} = \sum_{i \in \mathcal{T}} -\log(f(x)_i) \cdot f(x)_i. \quad (2)$$

- (*not-before; not-after*) interval instead of single point annotations
 - e.g. (1260; 1269)
- Modified **Huber loss** for intervals
- Pulls the output of the network more towards the middle of the interval



	Font Acc. ↑	Script Acc. ↑	Location Acc. ↑	Date MAE ↓
Textline system	98.42 %	88.54 %	69.85 %	21.91 years
Patch system	95.68 %	80.26 %	75.08 %	32.45 years
Linear fusion	98.27 %	88.84 %	70.77 %	21.99 years
Log-linear fusion	98.48 %	88.60 %	79.69 %	—
Baseline	—	55.22 %	62.46 %	—
The North LTU	82.80 %	74.12 %	43.69 %	79.43 years
CLUZH	95.66 %	35.25 %	—	—
NAVER Papago	97.17 %	—	—	—