

A Character Style Library for Syriac Manuscripts

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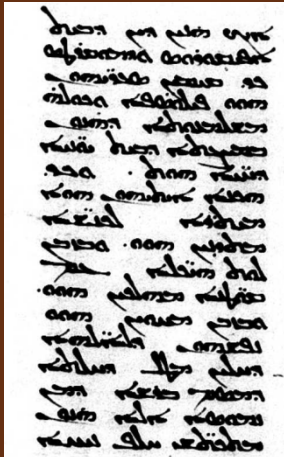


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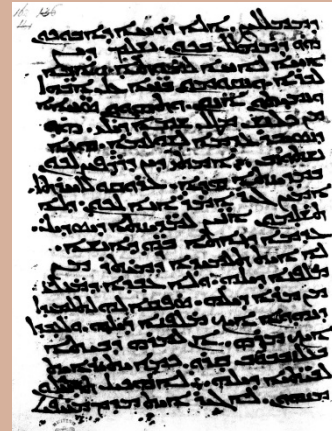


Paleography

- Most documents not securely dated
- Writing style changes over time
- Some documents have known dates
- Use these to calibrate dates of others



BL Add. 12150: 411 CE



BL Add. 14490: 1089 CE

300 CE

500 CE

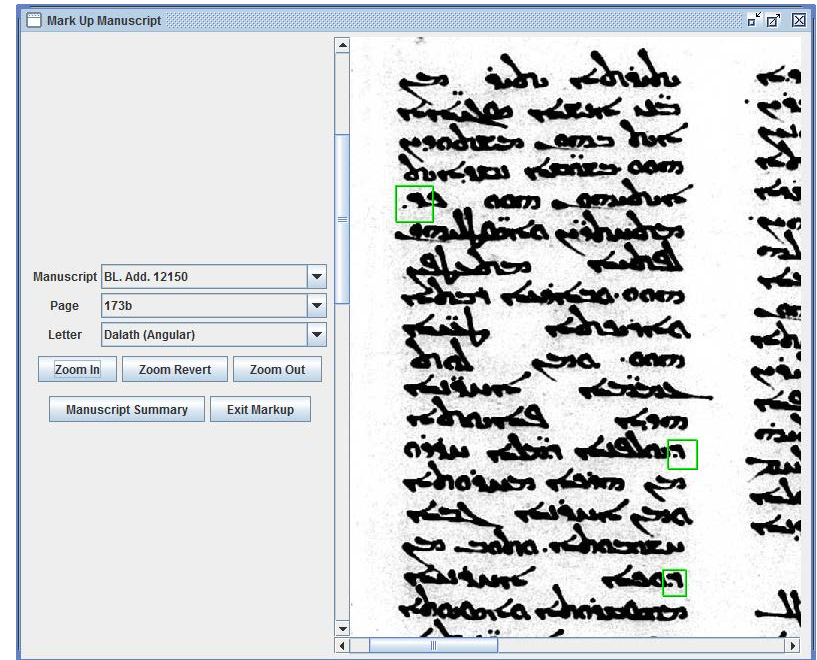
700 CE

900 CE

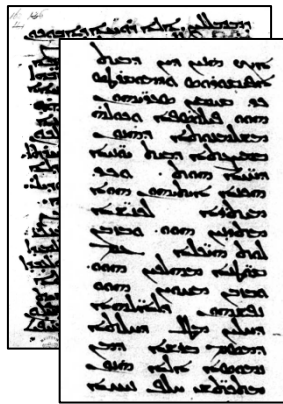
1100 CE

Human Annotation

- Humans identify character samples
- 5 per document per character
- Bounding box only
 - More detail too time-consuming
 - Need automatic character segmentation



Workflow Sketch



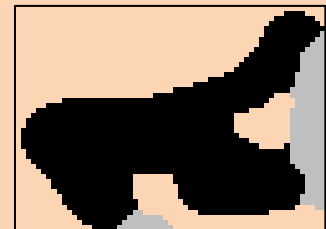
Document pages



Bounding boxes



Binarization



Character segmentation



Topological error detection



Further applications

Shaded tasks are carried out by computer

Computational Steps



1. Binarization

- Take heterogeneous sources to known format
- Uses Howe's binarization (Laplacian energy min.)

2. Character segmentation

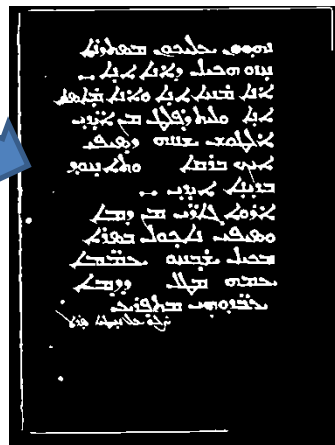
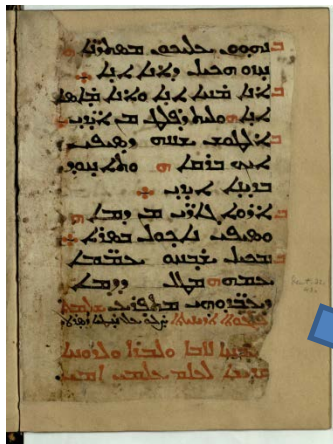
- Connected letters make problem tricky
- Evaluated two part-structured models

3. Postprocessing/quality control

- Possible to detect errors in prior stages

Binarization

- Most documents binarize well – automatic
- Two problem areas: red text & high resolution



Red text lost



Without smoothing

With smoothing

Part-Structured Models



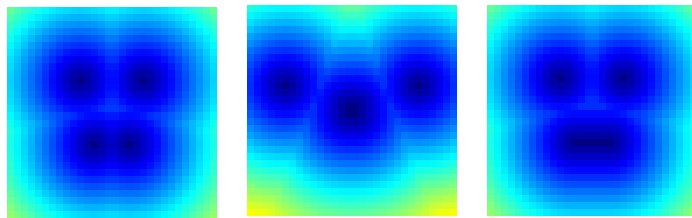
- Complex model is made of simple parts in a spatial relationship
- Proposed layout of parts is a **configuration**
- Likelihood of configuration has two factors:
 - Do observations support layout of parts? E_ω
 - Does layout of parts match expected offsets? E_ξ

The diagram illustrates the combination of two energy maps. On the left, a blue square labeled E_ξ is added to another blue square labeled E_ω . An arrow points to the resulting energy map E , which is a square with a color gradient from blue to yellow to red. To the right of the arrow, the equation $E = E_\xi + \lambda E_\omega$ is written.

$$E = E_\xi + \lambda E_\omega$$

Part-Structured Localization

- Part detectors do some localization



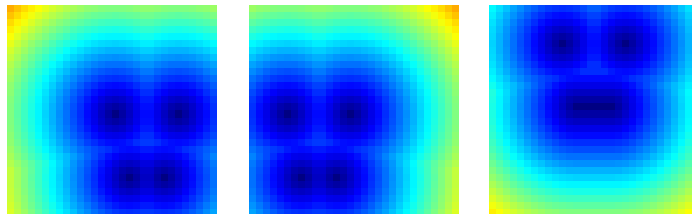
Eyes

Nose

Mouth



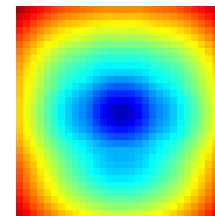
- Offset detections and combine



Left eye
to nose

Right eye
to nose

Mouth
to nose



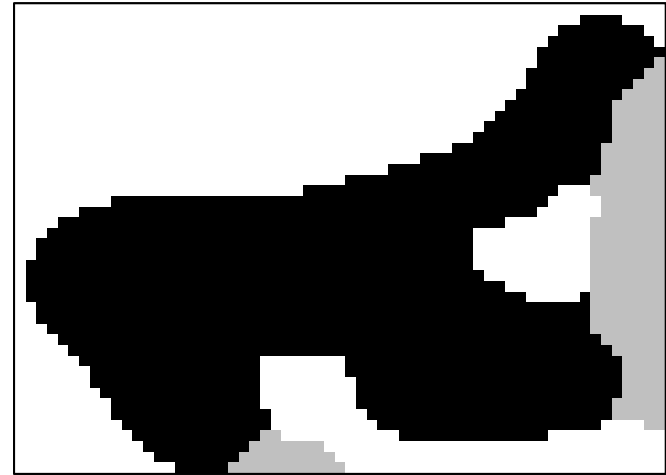
Combined
nose
likelihood

*Accounting for
subordinate
parts clarifies
nose position*

*Given nose
position, can place
subordinate parts*

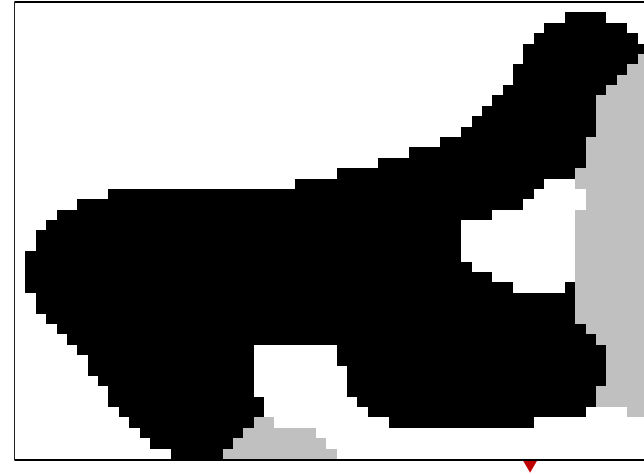
Model #1: Inkballs

- Parts are disks of ink placed on medial axis
- Model built from sample character
- Matching & segmentation:
 - Find minimal energy configuration
 - Render model to classify medial points
 - Attribute pixels based on nearest medial point



Model #2: Boundary Trace

- Parts are oriented edge segments
- Arranged in double ring around letter
- Matching & segmentation:
 - Find minimal energy
 - Identify closed loop
 - Attribute points



Similar to active contours/snakes except:

- *Prior on shape from model character*
- *Direct optimization*

Automatic Quality Control

- Topological considerations catch some errors
 - *Unimplemented: feedback to binarization step*



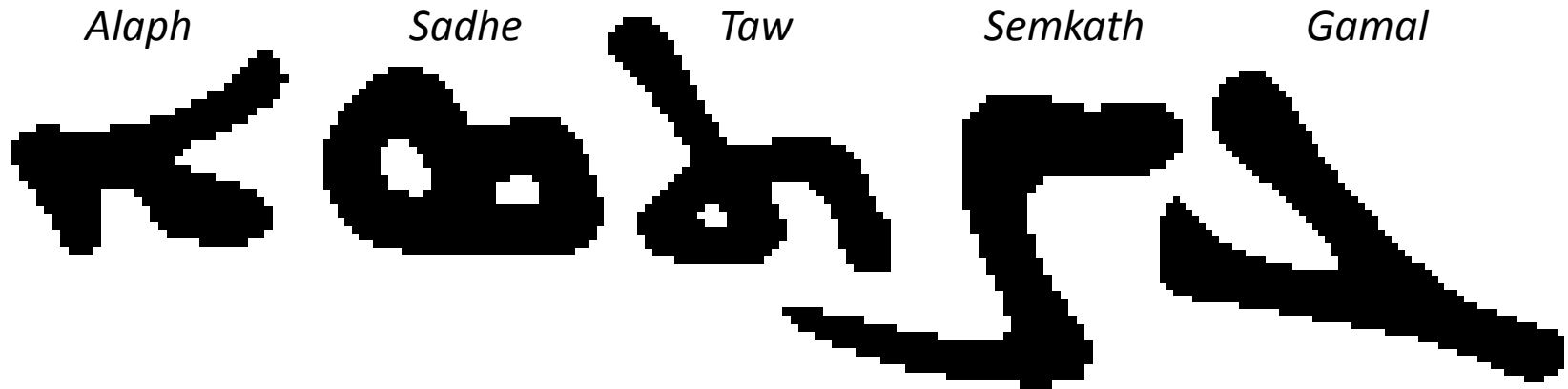
Filled holes



Broken characters

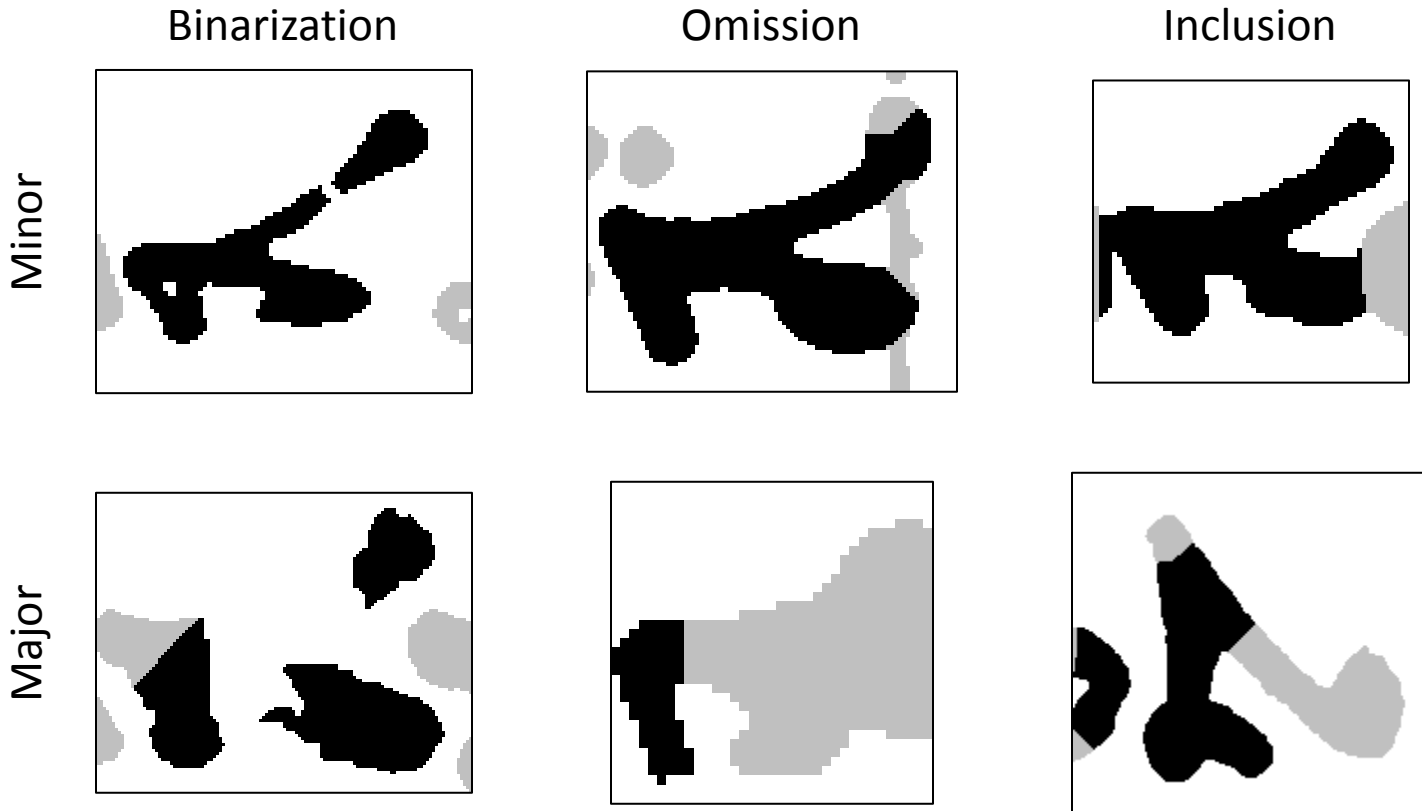
Manual Evaluation

- Evaluating 60,000 results is impractical
- Selected 1 sample per document x 5 letters

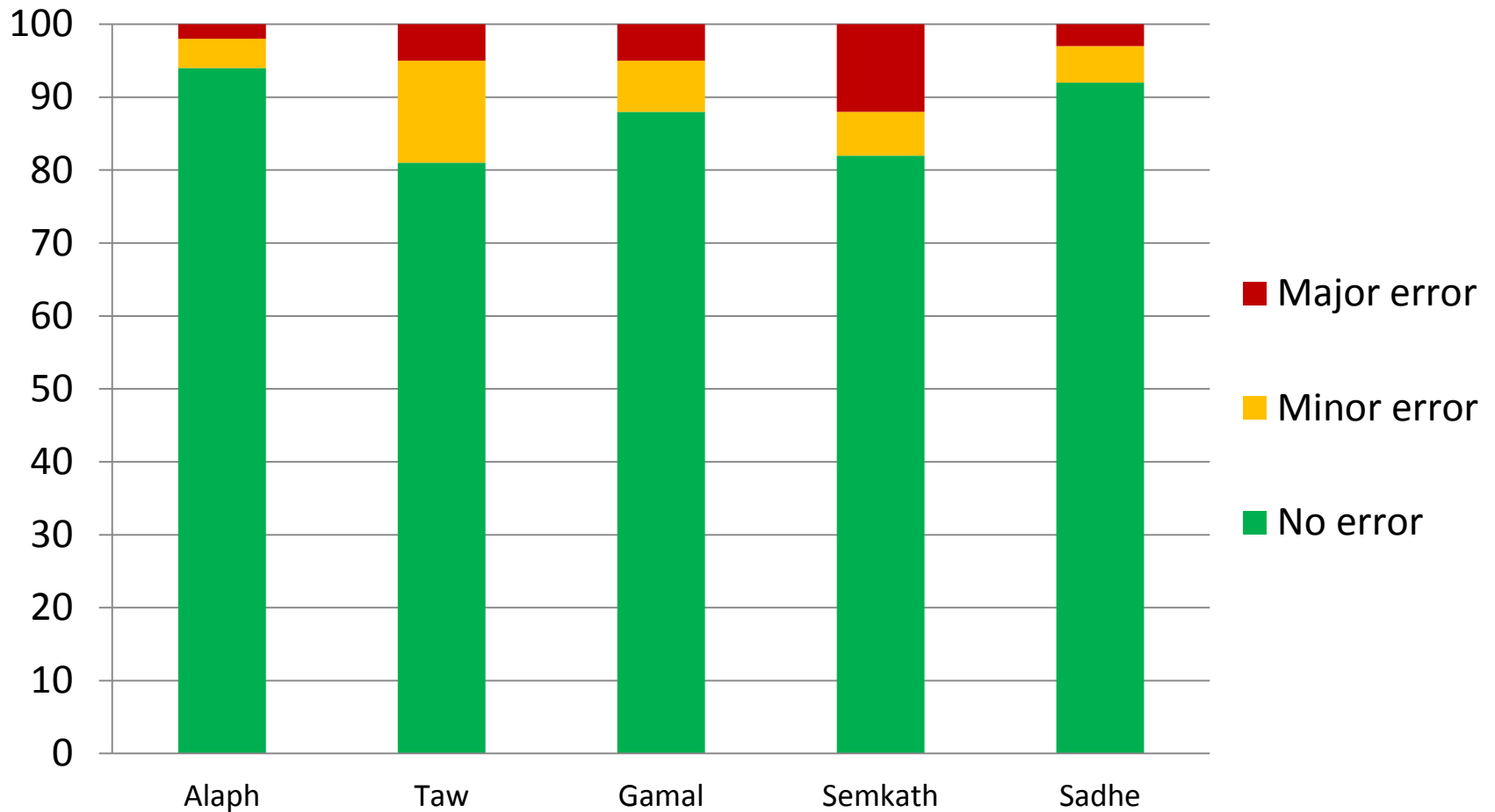


- Human expert rated quality of each result

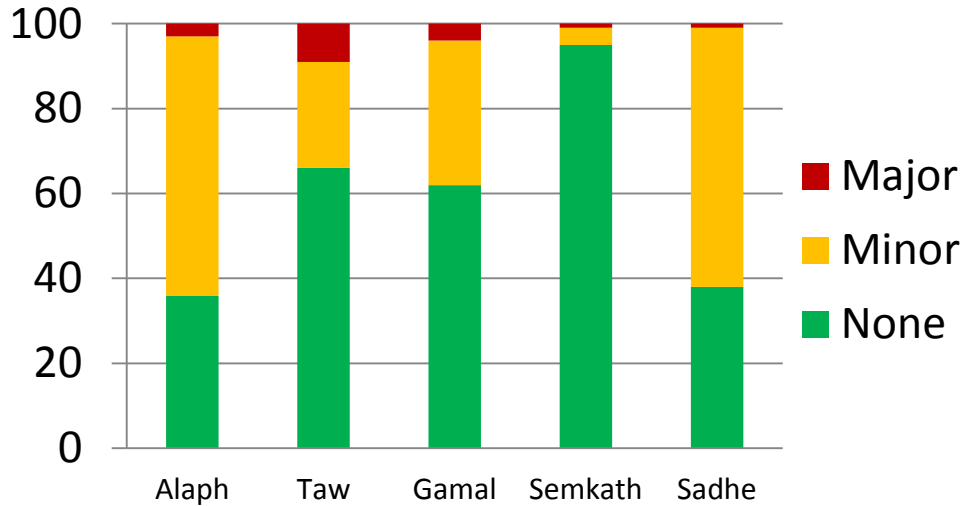
Error Types



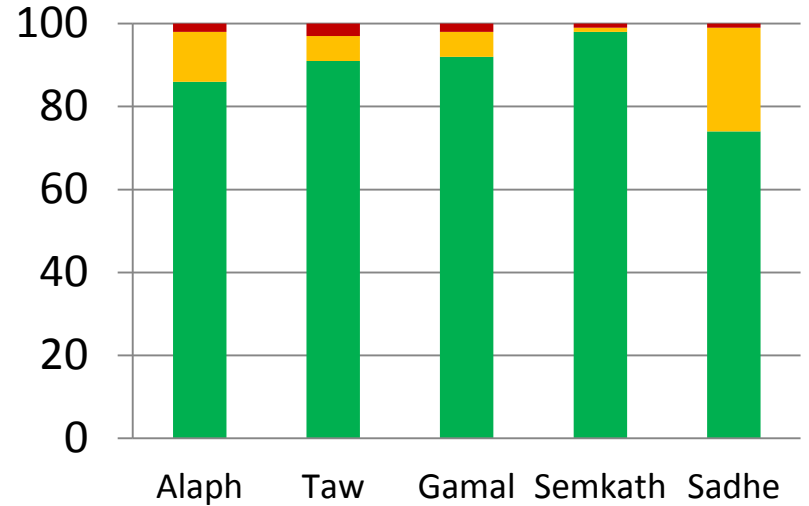
Results – Binarization Quality



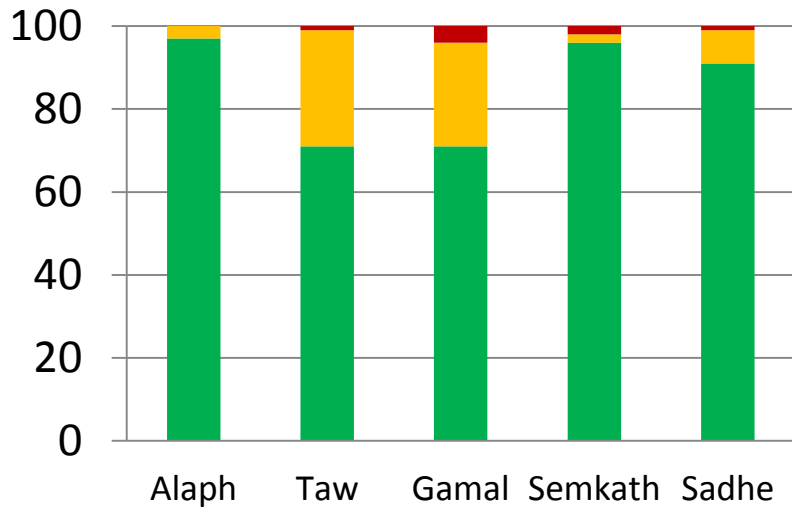
Omissions – Inkball



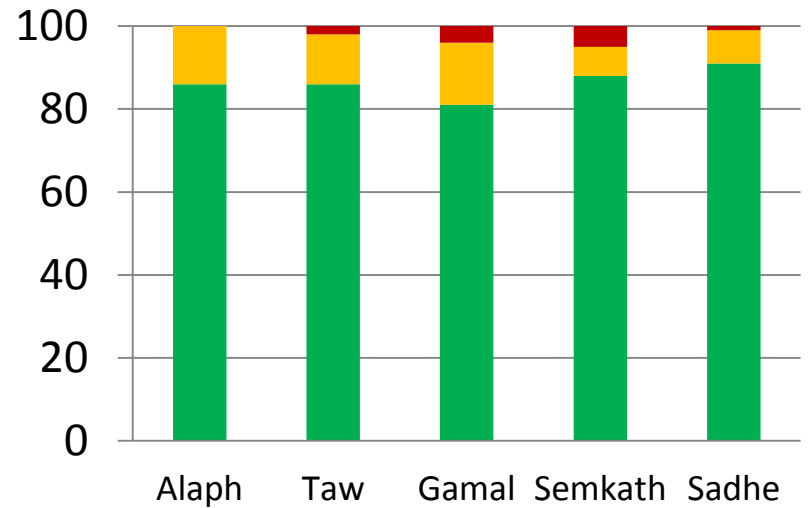
Omissions – Boundary



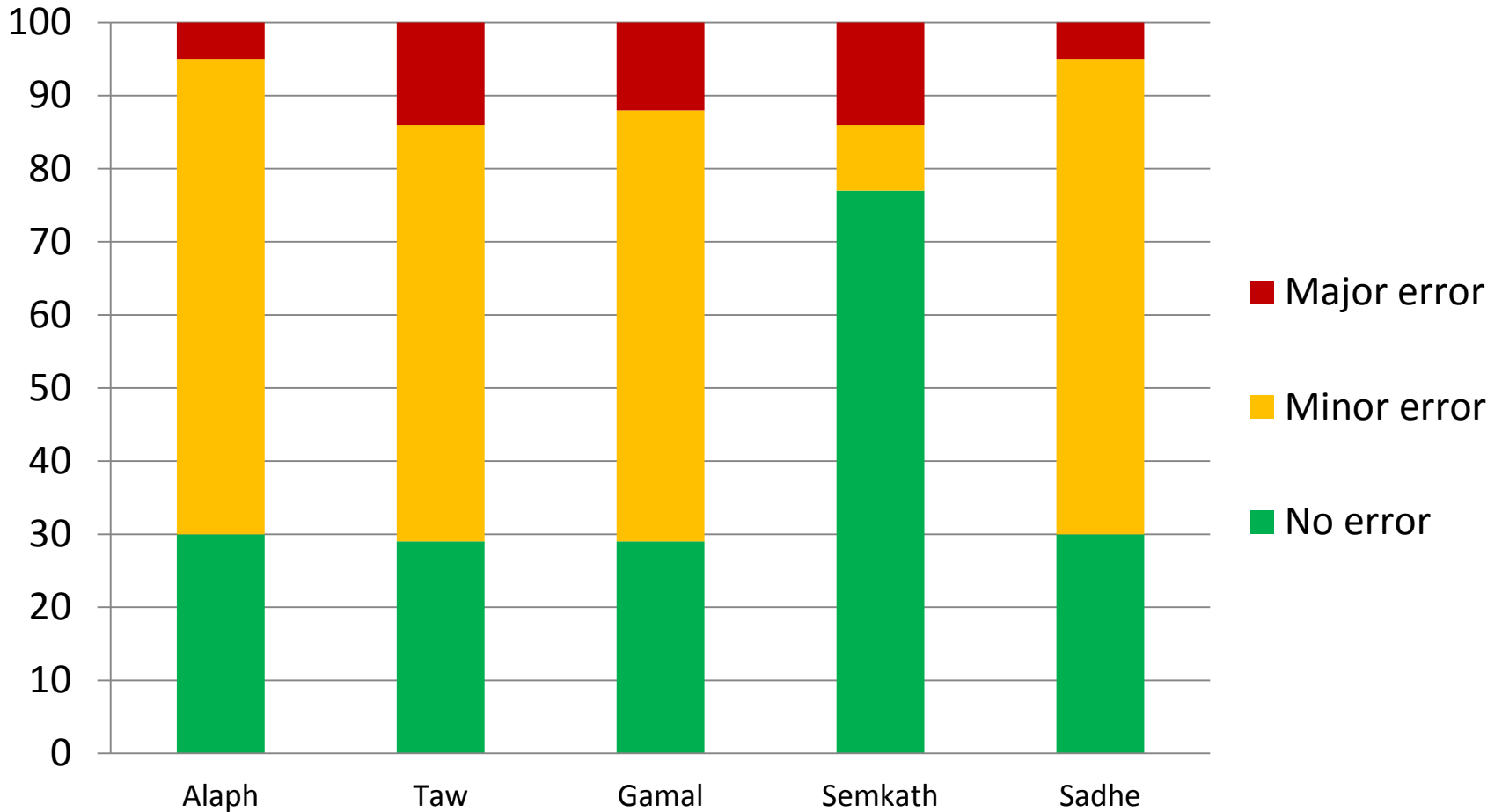
Inclusions – Inkball



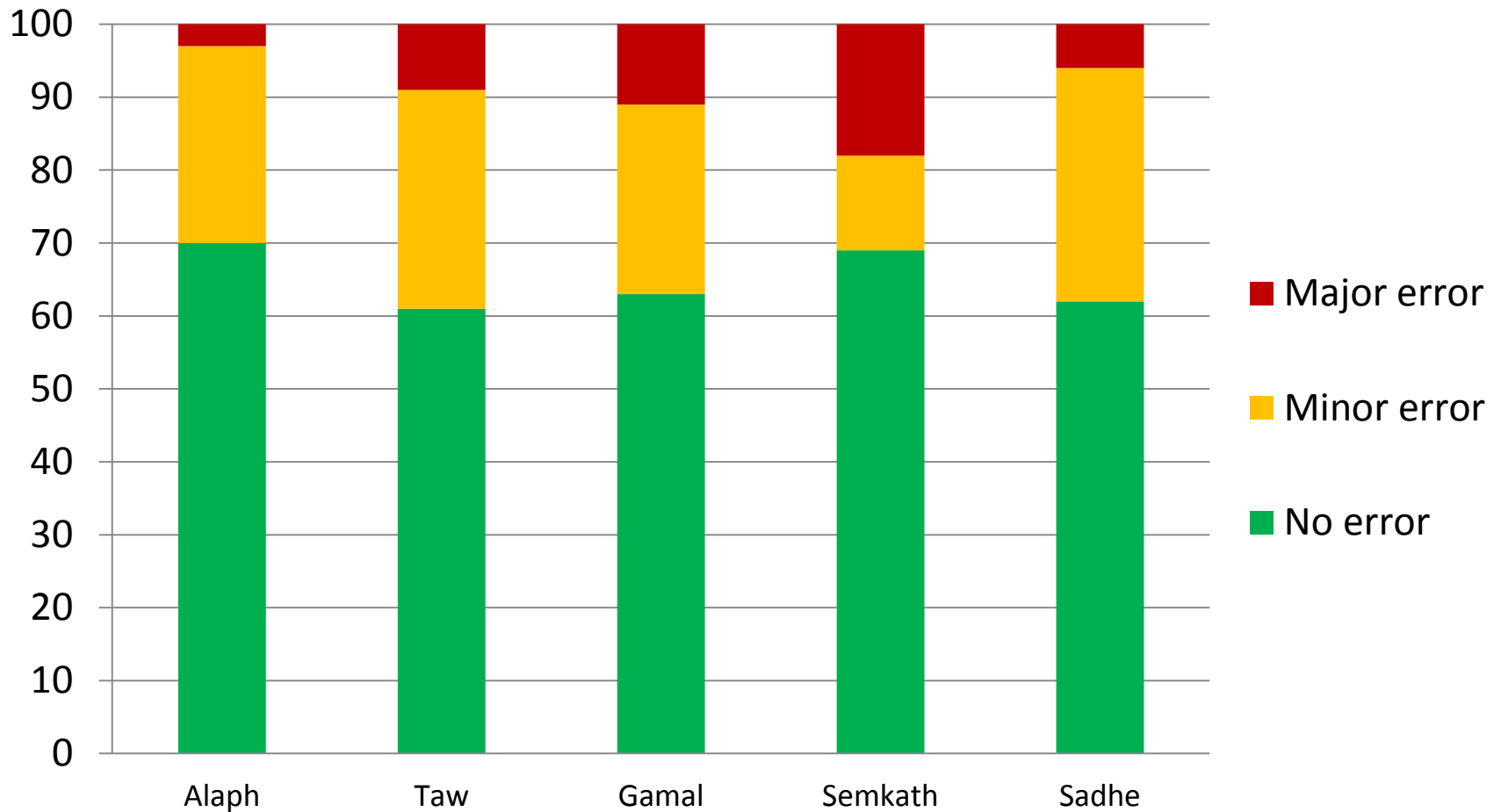
Inclusions – Boundary



Overall Results – Inkball



Overall Results – Boundary



Conclusion

- Mixture of human and machine effort



- Boundary models give best results:
 - At least 60% of samples are error-free
 - Fewer than 20% show major errors
 - 5 samples/character/document → likely one is good
- Next step: **paleographic dating**