

Writer Identification Using VLAD Encoded Contour-Zernike Moments

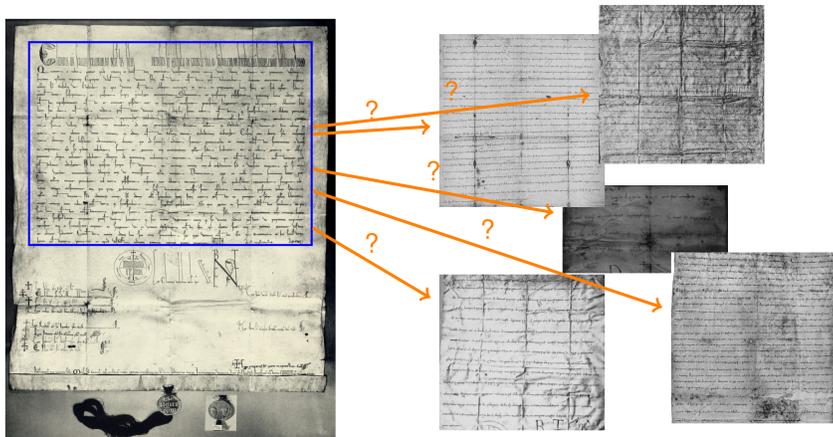
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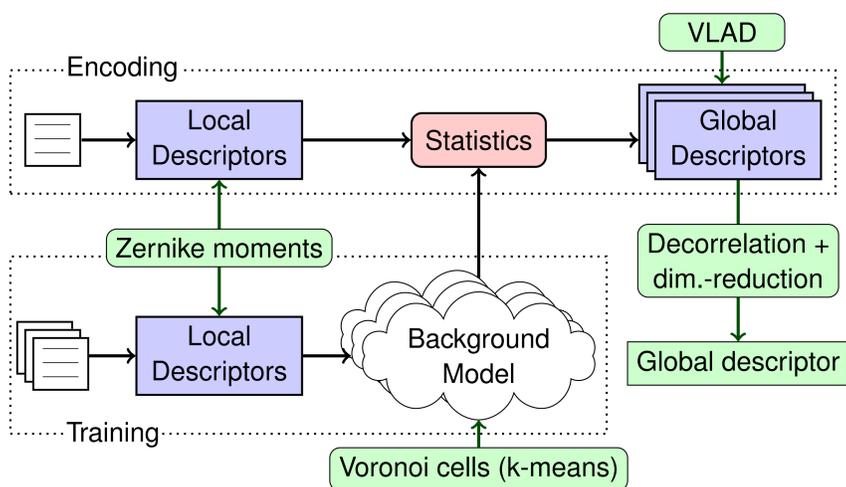
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Motivation



[Image source: Göttingen Academy of Sciences and Humanities]

Method Overview



Feature Extraction

- Evaluate Zernike moments [1] at the writing contour

$$A_{nm} = \frac{n+1}{\pi} \iint_{x^2+y^2 \leq w/2} f(x,y) V_{nm}^*(x,y) dx dy$$

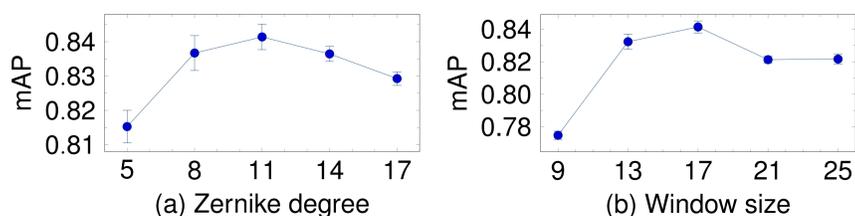
$$V_{nm}(\rho, \theta) = R_{nm}(\rho) e^{im\theta}$$

$$R_{nm}(\rho) = \sum_{s=0}^{(n-|m|)/2} \frac{(-1)^s [(n-s)!] \rho^{n-2s}}{s! (\frac{n+|m|}{2} - s)! (\frac{n-|m|}{2} - s)!}$$

$n \in \mathbb{Z}^+$ order $n \geq |m|$; $n - |m| = 2k, k \in \mathbb{Z}$
 $m \in \mathbb{Z}$ repetition
 $w \in \mathbb{Z}^+$ diameter

- Use A_{nm} directly \rightarrow rotation invariance not needed
- Use only positive $m \rightarrow$ 84 features instead of 156 (for degree 11)

Parameter Evaluation



Encoding

Vectors of locally aggregated descriptors (VLAD) [2]

Local Descriptors: $\mathbf{X} = \{\mathbf{x}_1, \dots, \mathbf{x}_T\}$; learned Dictionary: $\mathbf{D} = \{\mu_1, \dots, \mu_K\}$

$$\mathbf{v}_k = \sum_{\mathbf{x}_t: \text{NN}(\mathbf{x}_t=k)} (\mathbf{x}_t - \mu_k), \quad \text{NN: nearest neighbor of } \mathbf{x}_t \text{ in } \mathbf{D}$$

Postprocessing

- Intra-normalization
- Joint dimensionality reduction and whitening

Datasets

CVL

- 310 writers (training: 27, test: 283)
- 5 forms (1 German, 4 English)

Dann magst du mich in Fesseln schlagen,
Dann will ich gern zu Grunde geh'n!
Dann mag die Todtenglocke schallen,
Dann bist du meines Dienstes frey.

ICDAR13

- 350 writers (training: 100, test: 250)
- 4 forms (2 English / 2 Greek)

Τὴν εὐχρηστίαν τὰ εὐπείρα τὸν ἀφίπνου! Νὰ εὐχρὸς τὰ ἀφίπνου!
Νὰ ἀφίπνου εἶναι. Σὺ ποὺ τὰ φέρεται εὐχρὸς. Νὰ ἀφίπνου καὶ τὰ φέρεται.
Ὁμοίως εὐχρὸς ἀφίπνου! Τὴ εὐχρὸς ἀφίπνου; Νὰ φέρεται εὐχρὸς ἀφίπνου.

Evaluation

Hard criterion and mAP evaluated on ICDAR13 (test set)

	TOP-1	TOP-2	TOP-3	mAP
SV [3]	97.1	42.8	23.8	67.1
RootSIFT + VLAD + W.-full	96.1	51.7	29.1	70.7
SURF + VLAD + W.-full	95.6	50.6	28.2	70.5
Proposed	97.5	70.7	48.1	80.8
Proposed + W.-256	99.3	79.8	59.6	87.3
Proposed + W.-full	99.4	81.0	61.8	88.0

Hard criterion and mAP evaluated on CVL (test set)

	TOP-1	TOP-2	TOP-3	TOP-4	mAP
Comb. [4]	99.4	98.3	94.8	82.9	96.9
SV [3]	99.2	98.1	95.8	88.7	97.1
Proposed	98.8	97.6	95.3	86.2	96.0
Proposed + W.-256	99.2	98.7	97.5	92.5	97.8
Proposed + W.-full	99.4	98.9	97.4	92.7	97.9

Conclusion

- Outperforms s. o. t. a. on ICDAR13 and CVL
- Winning algorithm of the "ICDAR2015 Competition on Multi-script Writer Identification using QUWI Database"
- Using only a few components (256) is sufficient for comparable results \rightarrow enables large scale writer identification

References

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