Batik Classification using ConvNet and Transfer Learning

Research progress report

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IKO61181 Advance Image Processing

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Overview

• Batik image classification using ConvNet LeCun et al. [2015]

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- Accuracy comparison with:
 - 1. Convolutional stacked autoencoder (Menzata [2014])
 - 2. Direct SIFT descriptor matching (Willy et al. [2013])

Methodology

ConvNet by LeCun et al. [2015]

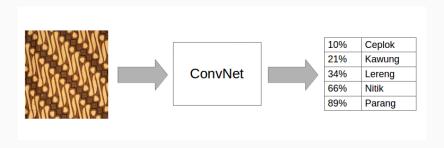


Figure 1: Classification with convolutional neural network

ConvNet by LeCun et al. [2015]

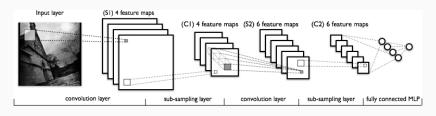


Figure 2: Convolutional neural network (source: deeplearning.net)

VGG16 by Simonyan and Zisserman [2014]

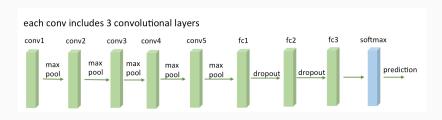


Figure 3: VGG-16 convnet (source: sebastianraschka.com)

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 - Initialize convnet weights with the weights of pretrained model (eg. VGG16 VGG19, ResNet50, Inception v3)

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- Using Python (OpenCV, Keras + Theano) https://github.com/ yohanesgultom/deep-learning-batik-classification

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- Run direct SIFT-descriptor matching on same data for comparison

References

- Yann LeCun, Yoshua Bengio, and Geoffrey Hinton. Deep learning. *Nature*, 521(7553):436–444, 2015.
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