# Landmarks Detection by Applying Deep Networks

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# Context



# Morphometry analysis

- Used to study the complex interaction between the evolution of insect and environmental factors.
- Characterize the common information of biological shape, such as, shape, sizes, or landmarks,....

#### Landmark

- A kind of point of interest
- ► A specific point defined by biologist. For example, intersection of viens on fly wing, the tip of beetle's mandible,...

# **Dataset**

- Images have been taken from 293 beetles, seperate into 5 parts (images),
- Format: 2D in RGB color.
- ► Focus on **pronotum** images.





(a) Left mandible



(b) Right mandible



(c) Body

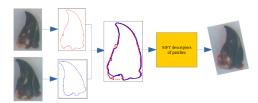


(d) Head

# **Problems**



## With segmentable images:1



<sup>&</sup>lt;sup>1</sup>Van-Linh Le, Marie Beurton-Aimar, Adrien Krähenbühl, and Nicolas Parisey. "MAELab: a framework to automatize landmark estimation." WSCG 2017.

## **Problems**



### With segmentable images:1



#### With un-segmentable images:



<sup>&</sup>lt;sup>1</sup>Van-Linh Le, Marie Beurton-Aimar, Adrien Krähenbühl, and Nicolas Parisey. "MAELab: a framework to automatize landmark estimation." WSCG 2017.

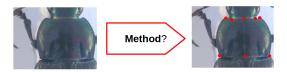
## **Problems**



#### With segmentable images:1



#### With un-segmentable images:



#### How to predict the landmarks coordinates?

<sup>&</sup>lt;sup>1</sup>Van-Linh Le, Marie Beurton-Aimar, Adrien Krähenbühl, and Nicolas Parisey. "MAELab: a framework to automatize landmark estimation." WSCG 2017.

# Content



#### Deep learning and Convolutional Neural Networks

Deep learning Convolutional neural networks (CNNs)

#### Proposed method

Network architectures Data augmentation Training

#### Experiments

#### Conclusion

# Deep learning



#### **Definition**

- A class of machine learning,
- Use a cascade of multiple layers for feature extraction and transformation,
- Learn multiple levels of representation in supervised or unsupervised.

# Deep learning



## Definition

- A class of machine learning,
- Use a cascade of multiple layers for feature extraction and transformation,
- Learn multiple levels of representation in supervised or unsupervised.

# **Applications**

- Computer vision (image recognition and classification)
- Speech recognition
- Question answering [], language translation[]

# **CNNs**

- Consists an input, an output and multiple hidden layers
- ► Arranges the data in 3 dimensions: width, height and depth
- Classical layers: convolutional layers (CONV), pooling layers (POOLING), dropout layers (DROPOUT), full-connected layers (FC), . . .

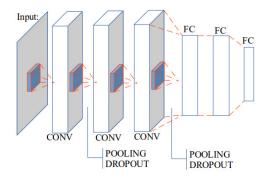


Figure: An example of CNN

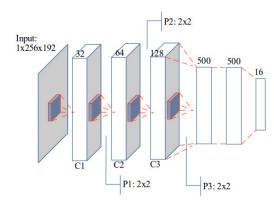


#### The first model includes:

- An gray-scale input,
- 3 CNN layers,
- 3 POOLING layers,
- ▶ 3 FC layers.

#### Problems:

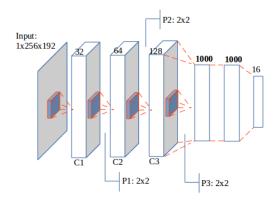
- Output is not good enough,
- Overfitting.





#### The second model:

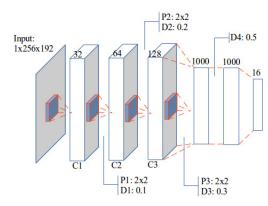
- Has the same architecture with the first one,
- Modify the output of FC layers,
- Result is not improved.





#### The third model includes:

- An gray-scale input,
- 3 CNN layers,
- ▶ 3 POOLING layers,
- 4 DROPOUT layers,
- ▶ 3 FC layers.



# Proposed method Data augmentation



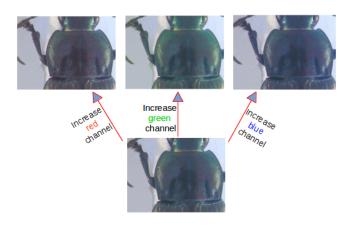
Dataset: 293 pronotum images in RGB format.

# Proposed method Data augmentation



### Augmentation methods:

Increase the value of each channel,

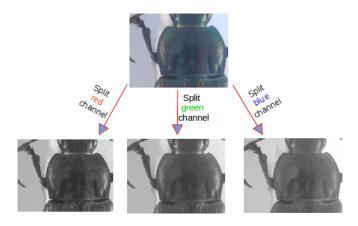


# Proposed method Data augmentation



#### Augmentation methods:

► Split the channels.



# Proposed method

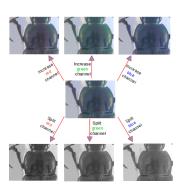
Data augmentation



Dataset: 293 pronotum images in RGB format. Augmentation methods:

- Increase the value of each channel,
- ▶ Split the channels.

Total:  $293 \times 7 = 2051$  images





- ▶ In Feather theme, the title page frame and the last frame have the Feather image as the background image.
- ► The Feather background image can be produced to any frame by wrating on the begining at the choosen frame the following

```
{\1bg
\begin{frame}[<options>]{Frame Title}{Frame Subtitle}
...
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