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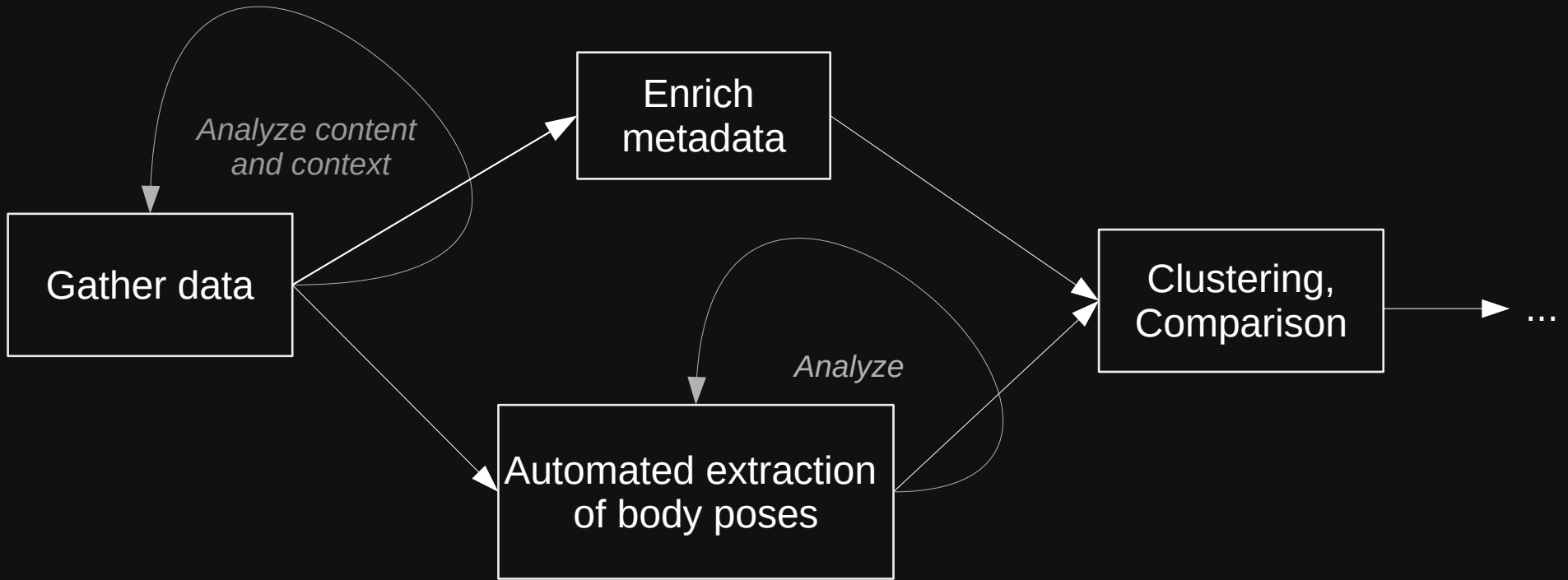
# Missing annotations: towards automated gesture recognition on artworks

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# Computational and Historical Analysis of Hands and Gestures in Early Modern Art

- Understanding most recurrent gestures in Early Modern art
- How gestures participate in the narrative system of paintings
- Who are the source painters of specific gestural patterns
- What are their evolutions

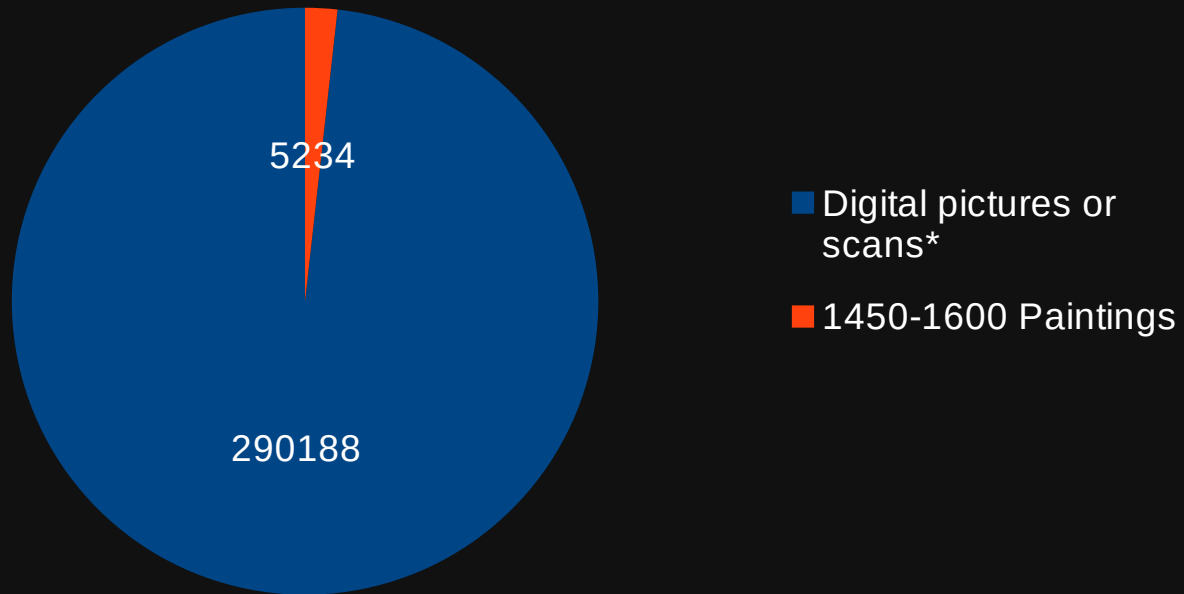
# First general steps



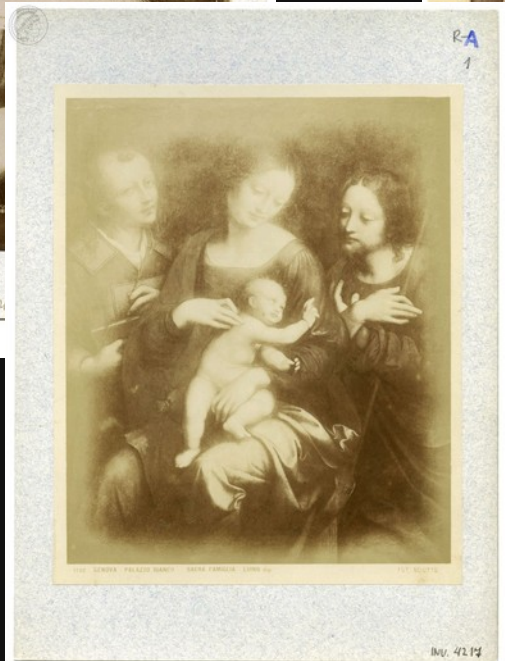
# The dataset

- Fototeca, collection of the Bibliotheca Hertziana

Objects present in the digital collection



# Visual content

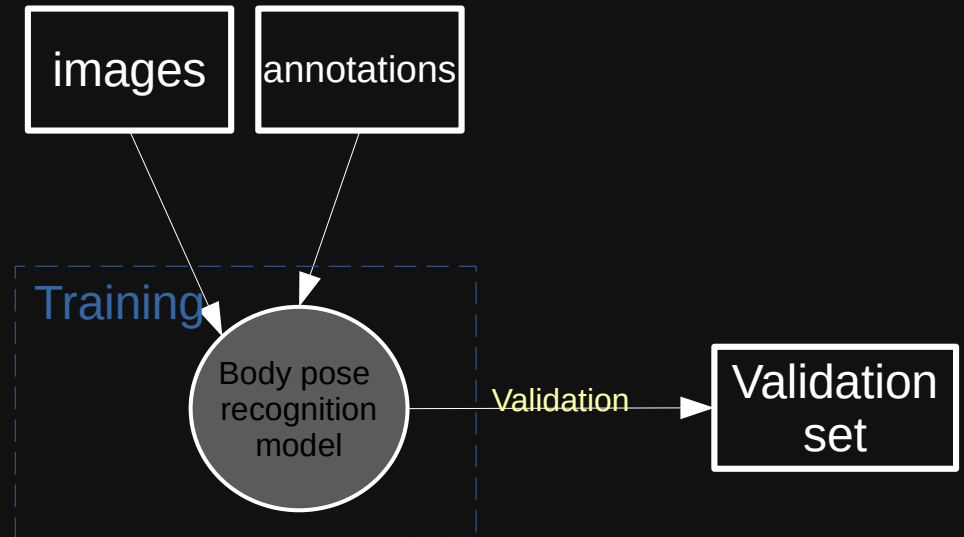


# Automated recognition of body poses

- How many bodies
- Where they are positioned on the picture
- Description of the body poses

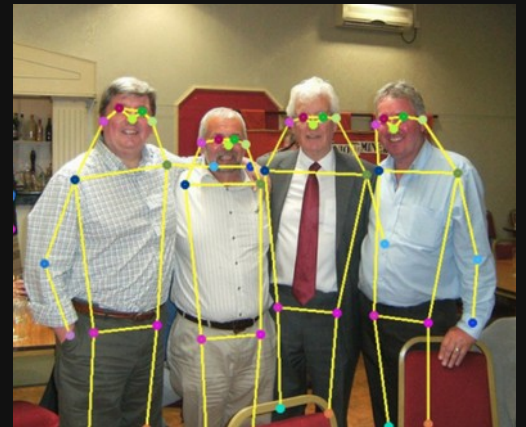
# Computer vision

- Human Pose Estimation
- Existing ML models: openpose, Detectron2...
- Big training datasets



# Microsoft COCO dataset

- 250'000 images of people with keypoints
- 17 body keypoints
- 68 face keypoints
- 42 hands keypoints





# Microsoft COCO dataset



# Problem

- Models trained on real images do not perform well on artworks
- Many differences of image content

# Main differences

- Layout
- Content
- Shape



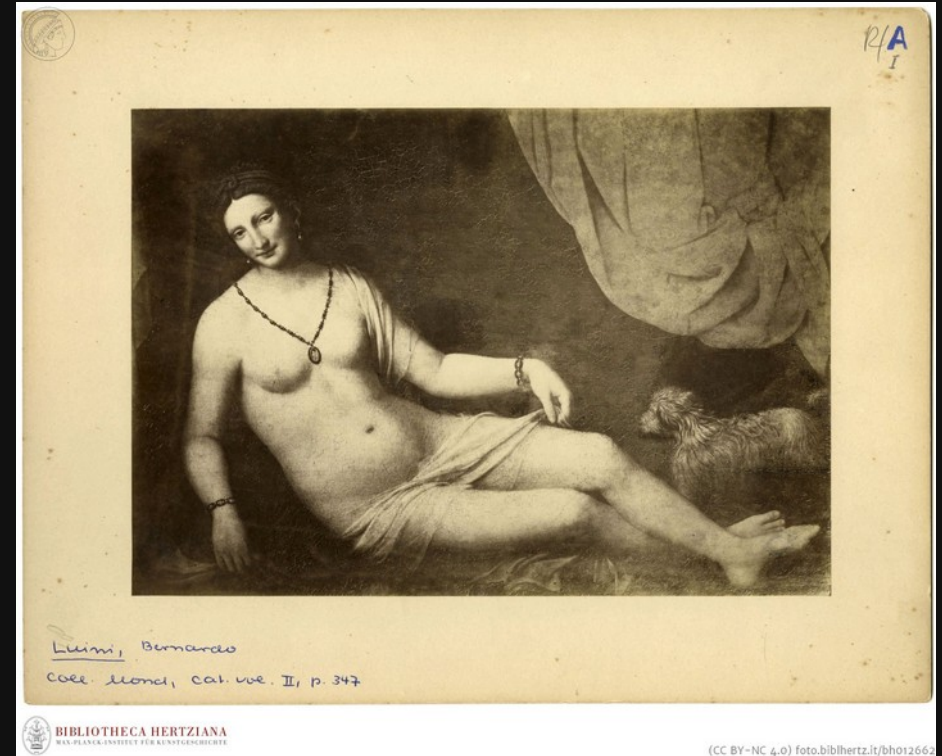
# Main differences

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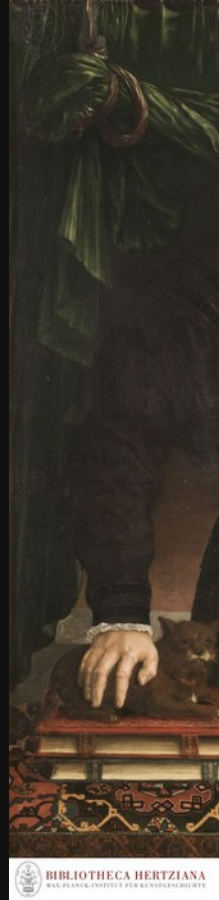
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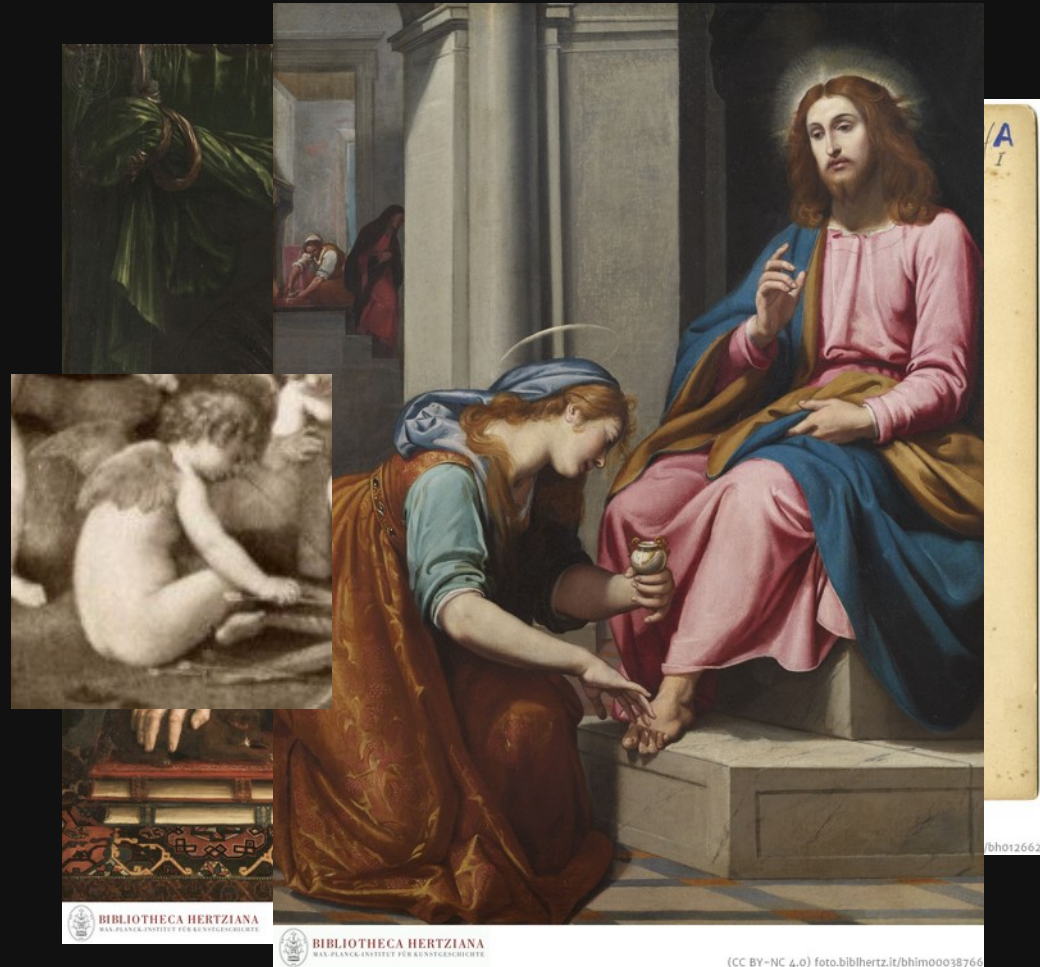
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# Main differences

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# Main differences

- Layout
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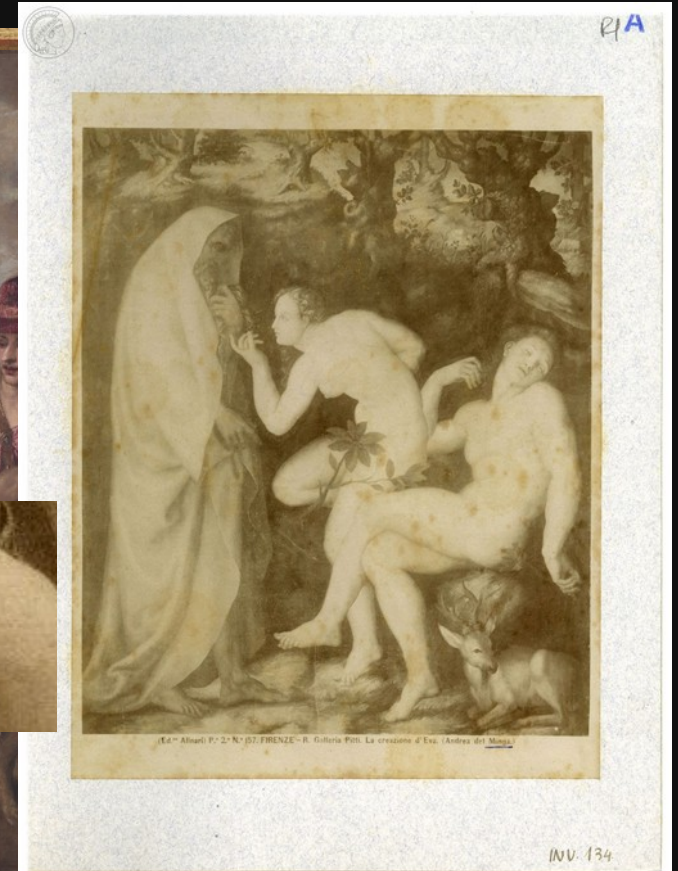
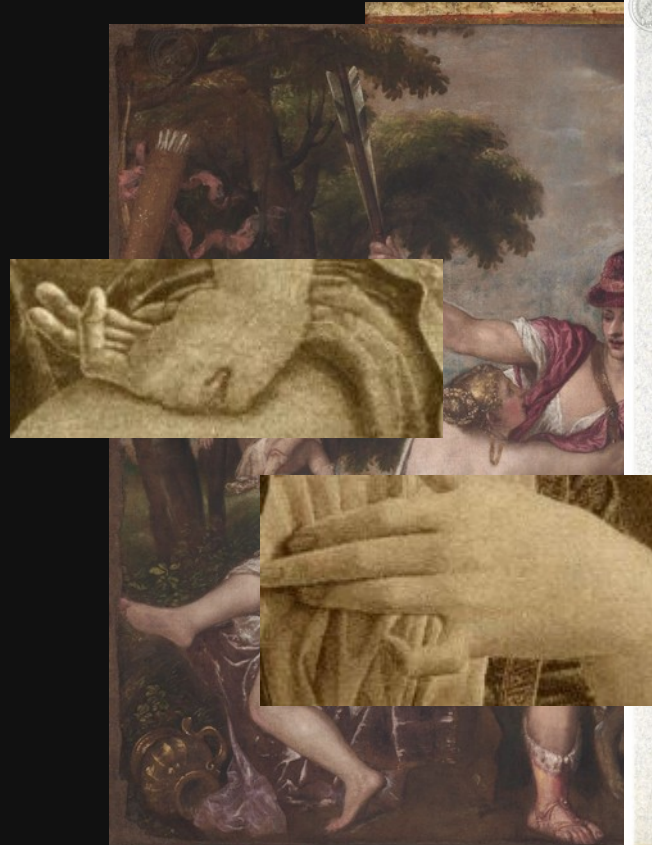
# Main differences

- Layout
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- Shapes

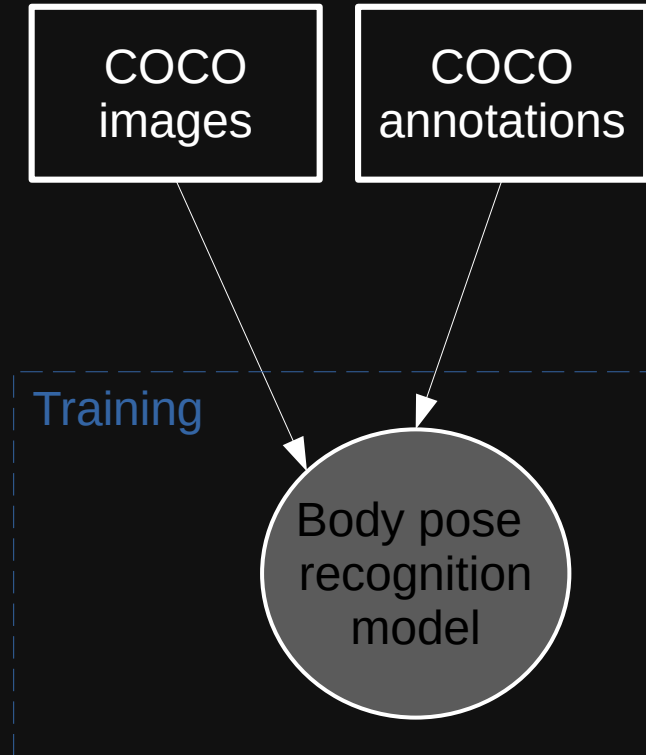


# Main differences

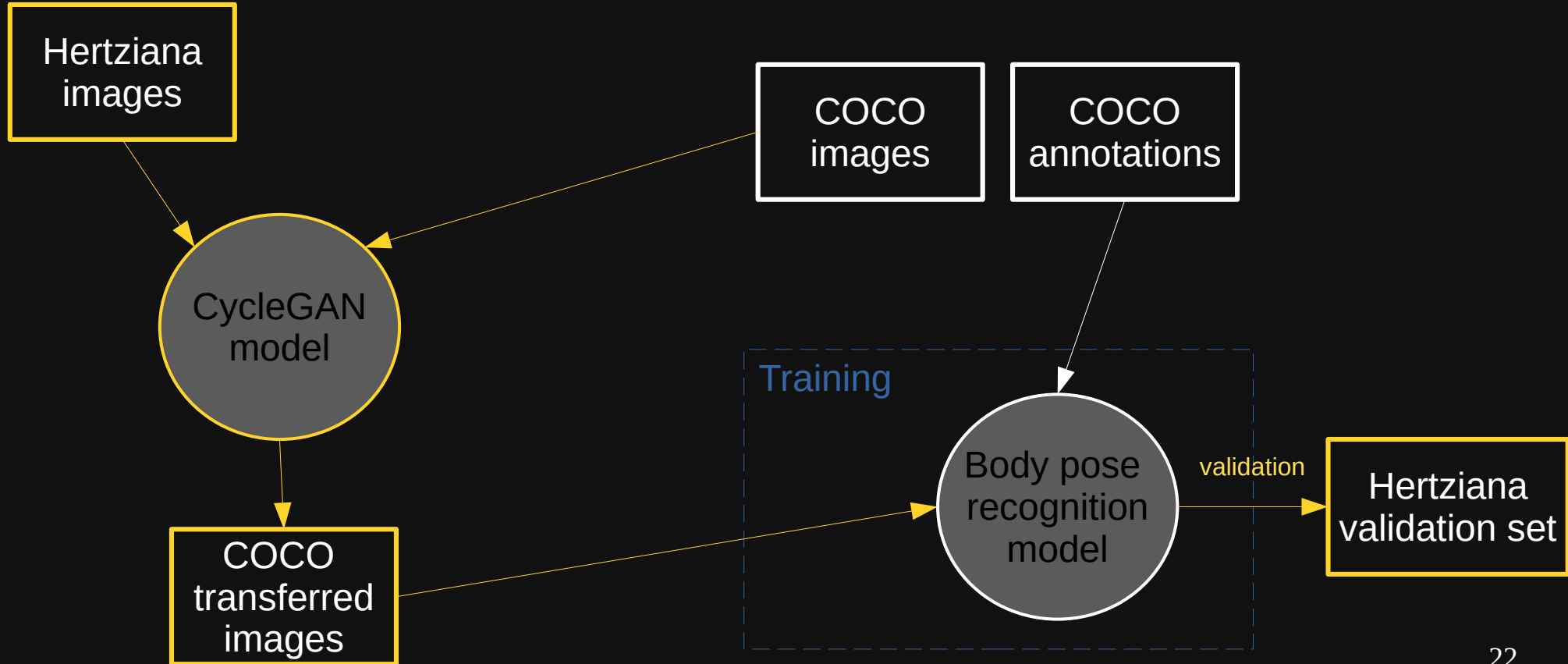
- Layout
- Content
- Shapes



# Domain adaptation

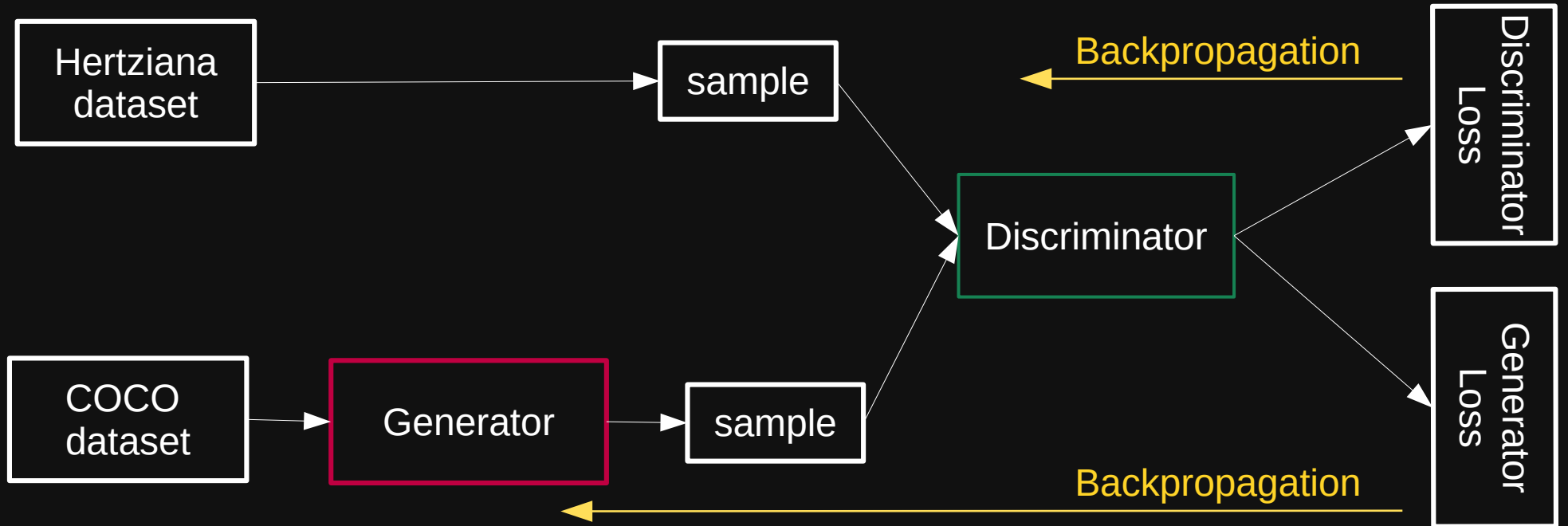


# Domain adaptation



# Domain adaptation

- Generative Adversarial Networks



# Domain adaptation

- Cycle GAN

Real image



Transferred image





# Domain adaptation

- Cycle GAN

Real image



Transferred image



# Domain adaptation

- Cycle GAN

Real image



Transferred image



# Domain adaptation

- Cycle GAN

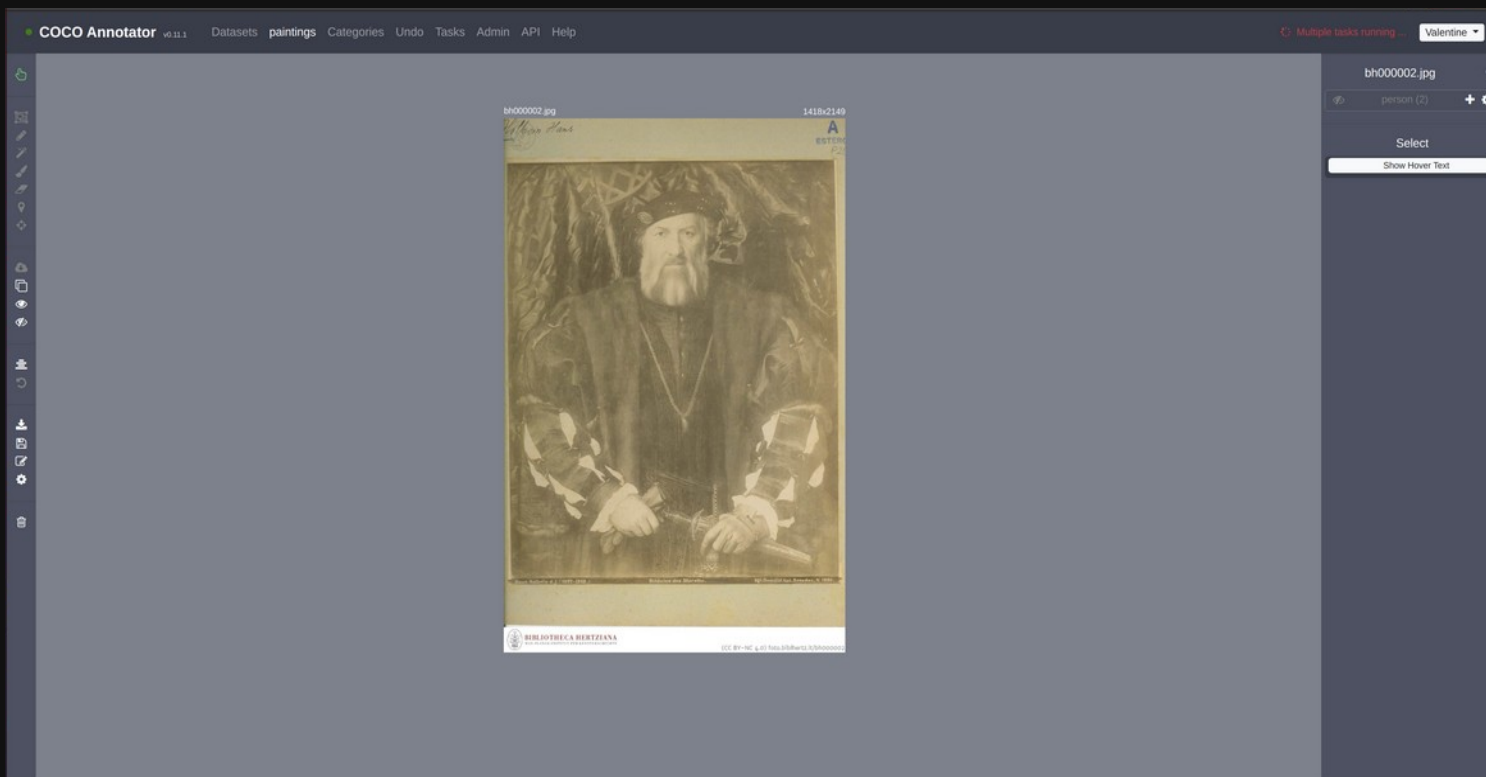
Real image



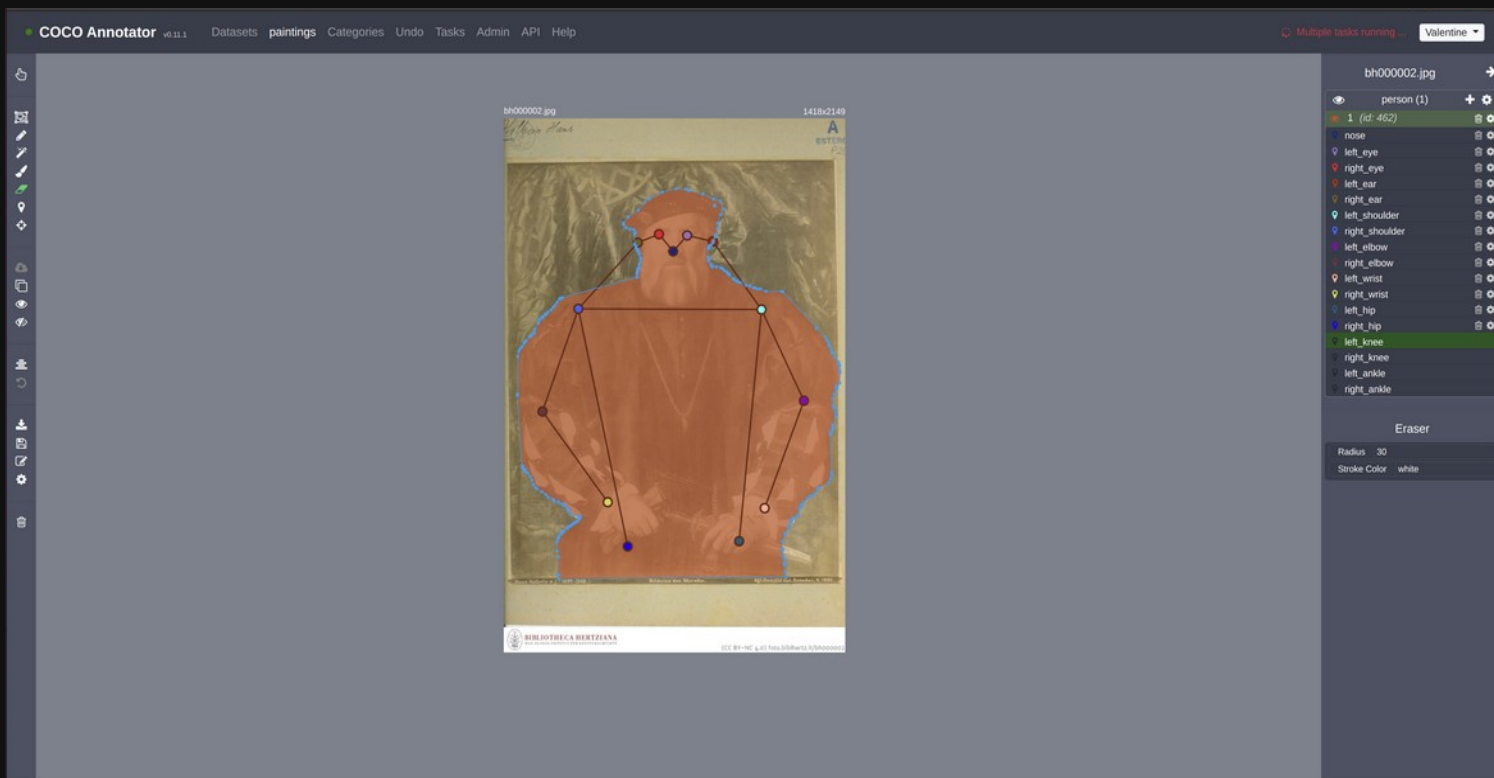
Transferred image



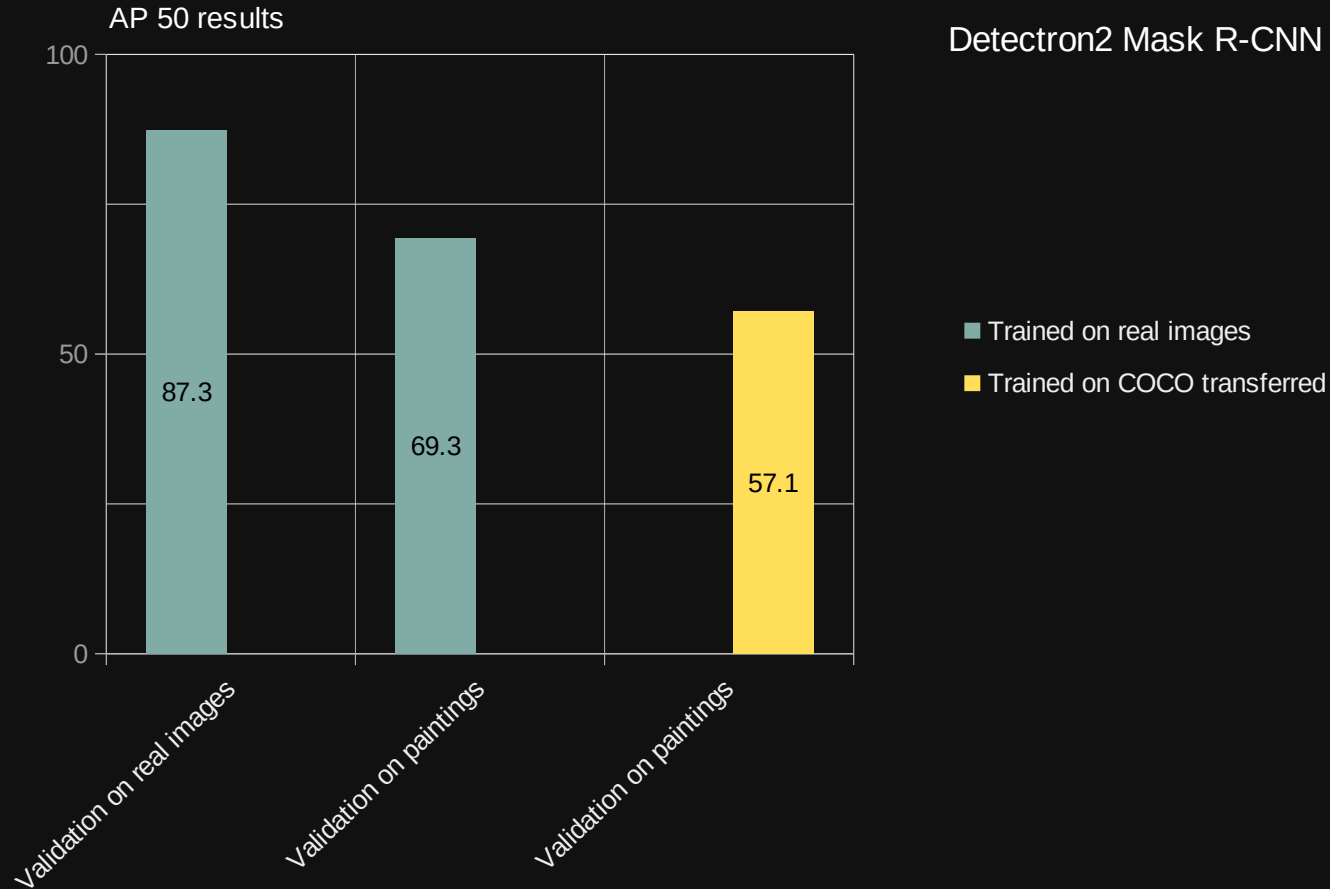
# Creation of a validation set



# Creation of a validation set



# First results comparison



# What else?

- Need for solutions to handle artistic specifics:
  - Context
  - Body morphology
  - Textures
  - Visual features
  - ...

# Pose Annotations Project for Artworks (PAPA)

- Creation of an annotated dataset for artworks
- Use standard COCO annotation format
- Participatory annotation platform:
  - Segmentation
  - Keypoints annotation
  - Validation



# Future questions

- Should we create more categories than *person*?
- To what kind of artwork could the annotations be extended?
- How do we annotate more abstract content?



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