

A review of the ChIA Project

Accessing and Analysing Cultural Images with New Technologies

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Project details

Funded in 2018 by the go!digital Next Generation call of the Austrian Academy of Sciences

Project duration: 24 months

Involved expertise

Digital Humanities, AI & NLP (ACDH-CH OeAW, AT)

Semantic technologies (Dublin City University, IE)

Cultural Image aggregation (Europeana Local - Österreich, AT)

Project team



PI: Amelie Dorn (ÖAW)



PI: Yalemisew Abgaz (IE)



Gerda Koch
(Europeana)



Renato Rocha Souza
(ÖAW)



Japesh Methuku
(IE)

with Ramiro Ortiz

Advisory board

- Artificial Intelligence: Ulla Kruhse-Lehtonen (Dain Studios) -FI
- Infrastructures and GLAM: Luca Pezzatti (E-RIHS) – IT
- Knowledge Design / DH: Jeffrey Schnapp – US
- Semantic Technologies: Anna Fensel (STI) – AT



Former Project members:

José Luis Preza Díaz (ÖAW)

Project team

Expertise



PI: Amelie Dorn (ÖAW)

DH, Cultural analysis



PI: Yalemisew Abgaz (IE)

Semantic technologies



Gerda Koch
(Europeana)

Image aggregation
Cultural Heritage



Renato Rocha Souza
(ÖAW)

AI, NLP, Machine learning



Japesh Methuku
(IE)

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Project aims



- engage and **test new technologies** (Semantic/ AI) against a background of a selected dataset of **food images**
- enhance **access** and **analysis** possibilities for **cultural data**



Image: Abraham van Beyeren, 1655, Mauritshuis. (CC-BY-PD)



Image: Abraham van Beyeren, 1655, Mauritshuis. (CC-BY-PD)

Title: Banquet Still Life



Image:
Balthasar van der Ast,
1620,
Mauritshuis.
(CC-BY-PD)



Image:
Balthasar van
der Ast,
1620,
Mauritshuis.
(CC-BY-PD)

Title: Fruit Still Life with Shells and Tulip

The major research questions

- How can we **explicitly** and **semantically represent** and **interlink** the rich information contained in **historical food images**?
- How can we support efficient search, analysis and exploit historical images by both **humans** and **machines**?
- What **AI tools** are available and how can we build AI tools for the exploitation of historical images?



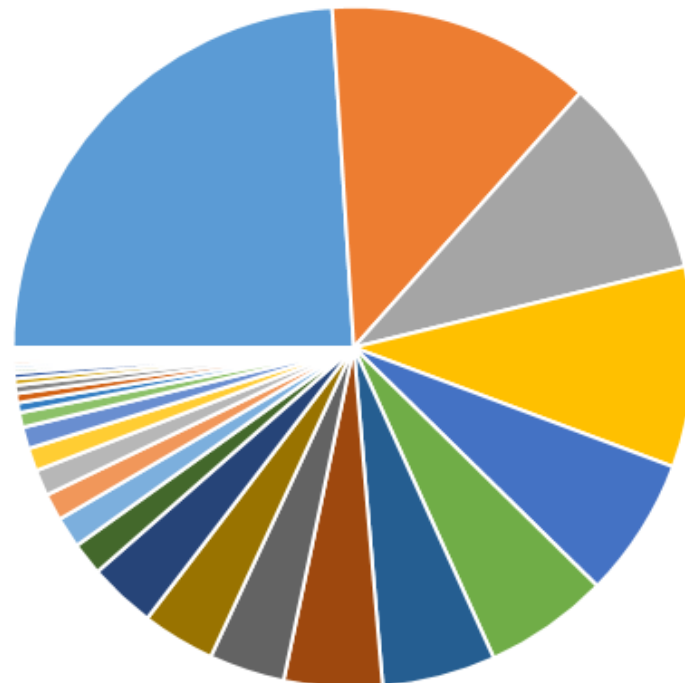
Image: Banquet Still Life (Adriaen Van Utrecht); CC-BY-PD

Europeana data set

Total: 58.6 Mio digital objects

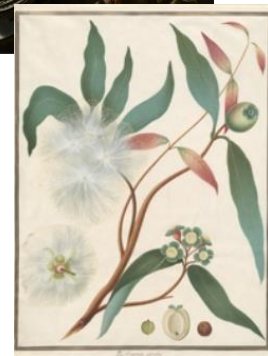
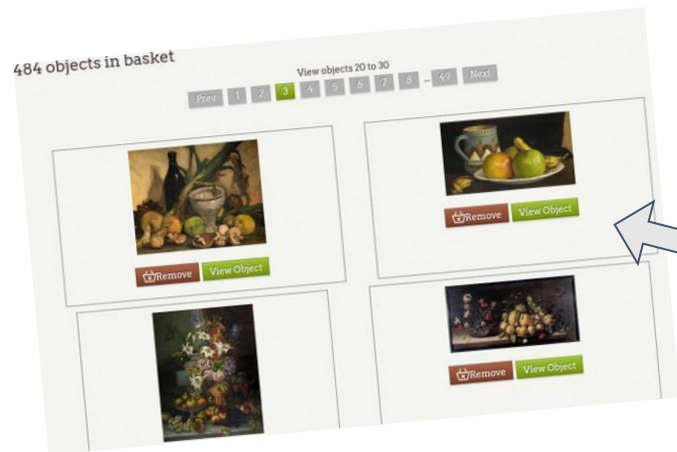
Includes: 34.2 Mio digital images

from: 3.500 institutions in 42 countries

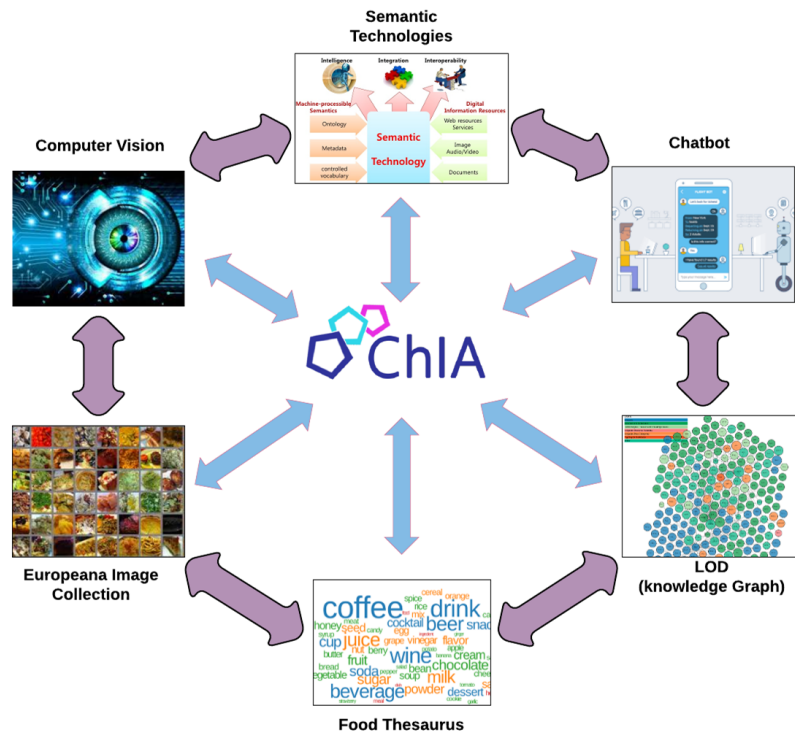


The ChIA dataset

- Selection based on food context of images
- 42.969 images (available with Free Access licenses); ~20.000 images dealing with “food” selected in form of various sets (baskets) for later download & analysis of metadata and images



Methods & Tools



- Semantic Technologies
- Chatbot Technology
- Knowledge Maps/Graphs
- Visual Search

The case of building an experimental dataset

- Bridging the gap between the information packed in the images and the explicit annotation of the content of the images using ontologies.
- Interactions between the team members to understand the problem and to work towards the solution

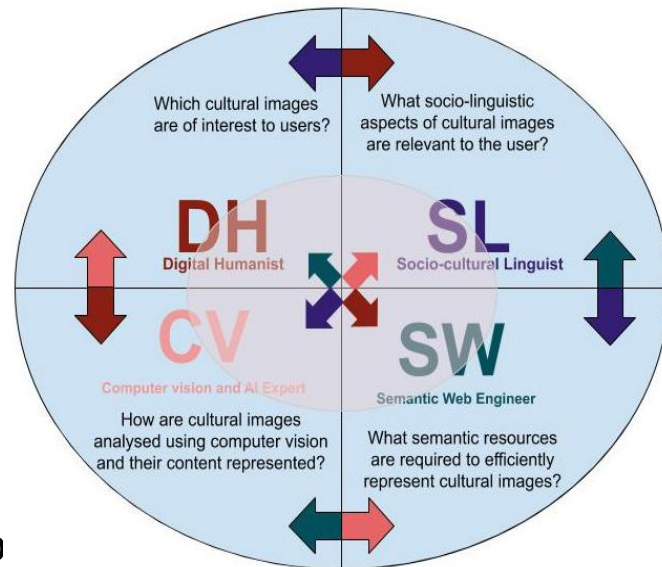


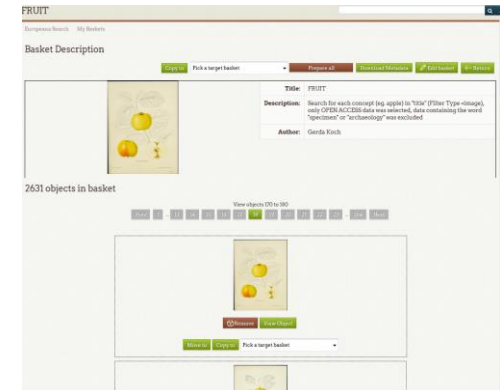
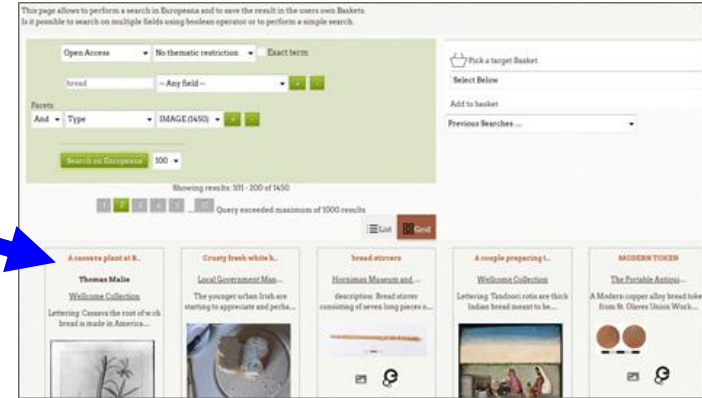
Image: CC-BY-4.0 Yalemisew Abgaz in Abgaz, Dorn, Koch & Preza Diaz. (2020).

Overall Outcomes

- the ChIA system
- A **search and exploration system** for Europeana datasets
 - experimentation with alternative modes of navigation
 - approach to objects within networks of relations
- **Reports** on advantages/challenges of the application of current and next technologies on the example of Europeana data

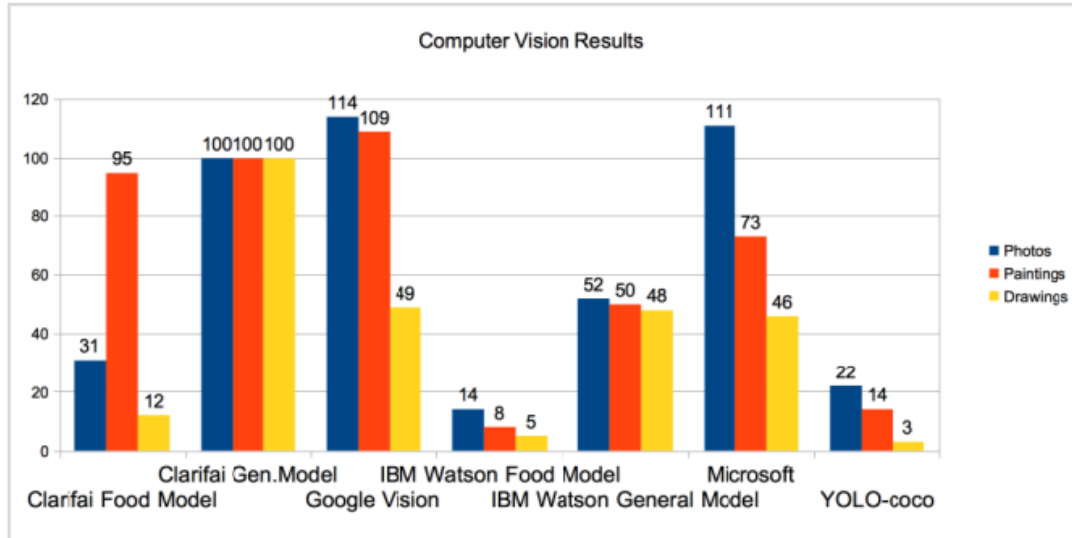
The ChIA intermediate infrastructure...

- ...was set up as one-stop shop for **access&download** of Europeana images and supports download of digital **images&metadata** in one process including a checking routine on data availability and data access rights
- ...provides for researchers the possibility to easily **generate** out of the wealth of (open access) Europeana digital content **customized test data sets** for further analysis with CV/CNN/AI tools.



Results: Computer Vision

Pilot-test on selected images (n=15) of different commercial (Google Vision, Clarifai, IBM Watson, Microsoft Services) and open-source (YOLO) Computer Vision (CV) tools for cultural food image analysis. 3 image categories: photographs, drawings, sketches



Source: Preza Diaz et al. 2020

- Not only quantity, but also quality of generated CV concepts seems important for successfully enriching cultural food images.
- Some types of images (e.g. sketches) particularly challenging to process for CV solutions.

Image classification

Assessing the (human) inter-annotator agreement

Task_1	Amelie	Gerda	Marcos	Renato	Yalemisew
Amelie	1.000/(392)	0.928/(392)	0.892/(392)	0.907/(392)	0.886/(391)
Gerda	0.928/(392)	1.000/(392)	0.892/(392)	0.938/(392)	0.896/(391)
Marcos	0.892/(392)	0.892/(392)	1.000/(392)	0.923/(392)	0.923/(391)
Renato	0.907/(392)	0.938/(392)	0.923/(392)	1.000/(392)	0.918/(391)
Yalemisew	0.886/(391)	0.896/(391)	0.923/(391)	0.918/(391)	1.000/(391)

Task_2	Amelie	Gerda	Marcos	Renato	Yalemisew
Amelie	1.000/(392)	0.330/(392)	0.252/(392)	0.316/(392)	-0.091/(392)
Gerda	0.330/(392)	1.000/(392)	0.210/(392)	0.306/(392)	0.153/(392)
Marcos	0.252/(392)	0.210/(392)	1.000/(392)	0.051/(392)	-0.031/(392)
Renato	0.316/(392)	0.306/(392)	0.051/(392)	1.000/(392)	-0.028/(392)
Yalemisew	-0.091/(392)	0.153/(392)	-0.031/(392)	-0.028/(392)	1.000/(392)

Task_3	Amelie	Gerda	Marcos	Renato	Yalemisew
Amelie	1.000/(392)	0.659/(392)	0.296/(392)	0.534/(392)	0.317/(392)
Gerda	0.659/(392)	1.000/(392)	0.325/(392)	0.453/(392)	0.268/(392)
Marcos	0.296/(392)	0.325/(392)	1.000/(392)	0.424/(392)	0.370/(392)
Renato	0.534/(392)	0.453/(392)	0.424/(392)	1.000/(392)	0.454/(392)
Yalemisew	0.317/(392)	0.268/(392)	0.370/(392)	0.454/(392)	1.000/(392)

Table and analysis @ Renato Rocha Souza

Study of available CNN architecture candidates for transfer learning

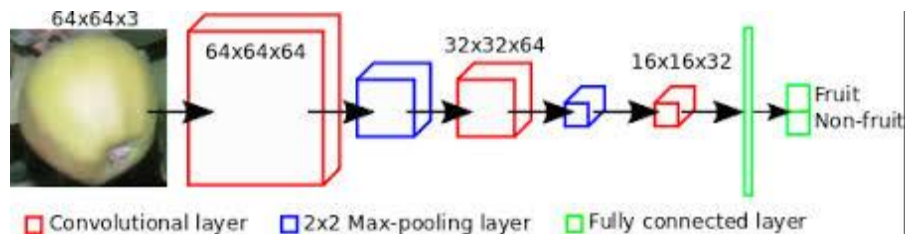


Image Classification

- Europeana aggregates millions of cultural objects including cultural images on its platform.
- Among the cultural images, we focus on food related images - We love culture and we also love food!
- The collection contains varieties of food images
- We wanted to answer the following questions
 - Can I find food images that contain fruit?
 - Can I find food images that are appealing?
 - Can I find food images that are formal?



The problem

- The answer to the previous questions is “May be”
- The main reasons are:
 - Cultural concepts such as “appealing” and “formal” are often difficult to understand and define
 - Not sufficient metadata/description is available
 - Existing computer vision is not yet effective in classifying cultural images

Methods

Our proposed method focuses on

- Formalisation: we use domain specific ontology terms from Existing vocabularies such as Getty Arts and Architecture Thesaurus , Iconclass and FoodOn Ontology
- Annotation: we annotate the images with three vocabulary terms
 - fruit/non-fruit Relatively less complex
 - appealing/non-appealing abstract and dependent on cultural background
 - formal/informal abstract and dependent on cultural background
- Model: we will train and build a CNN model using manually collected annotation
- Automatic annotation: Using the model, we will apply our solution to the bigger Europeana collection

Image Classification Task

- 1) Definition of a training dataset:
Still life images
- 1) Definition of classification protocol
- 2) Definition of food tags
& cultural tags
- 1) Deployment of tool (MakeSense.AI)



Image Classification Task

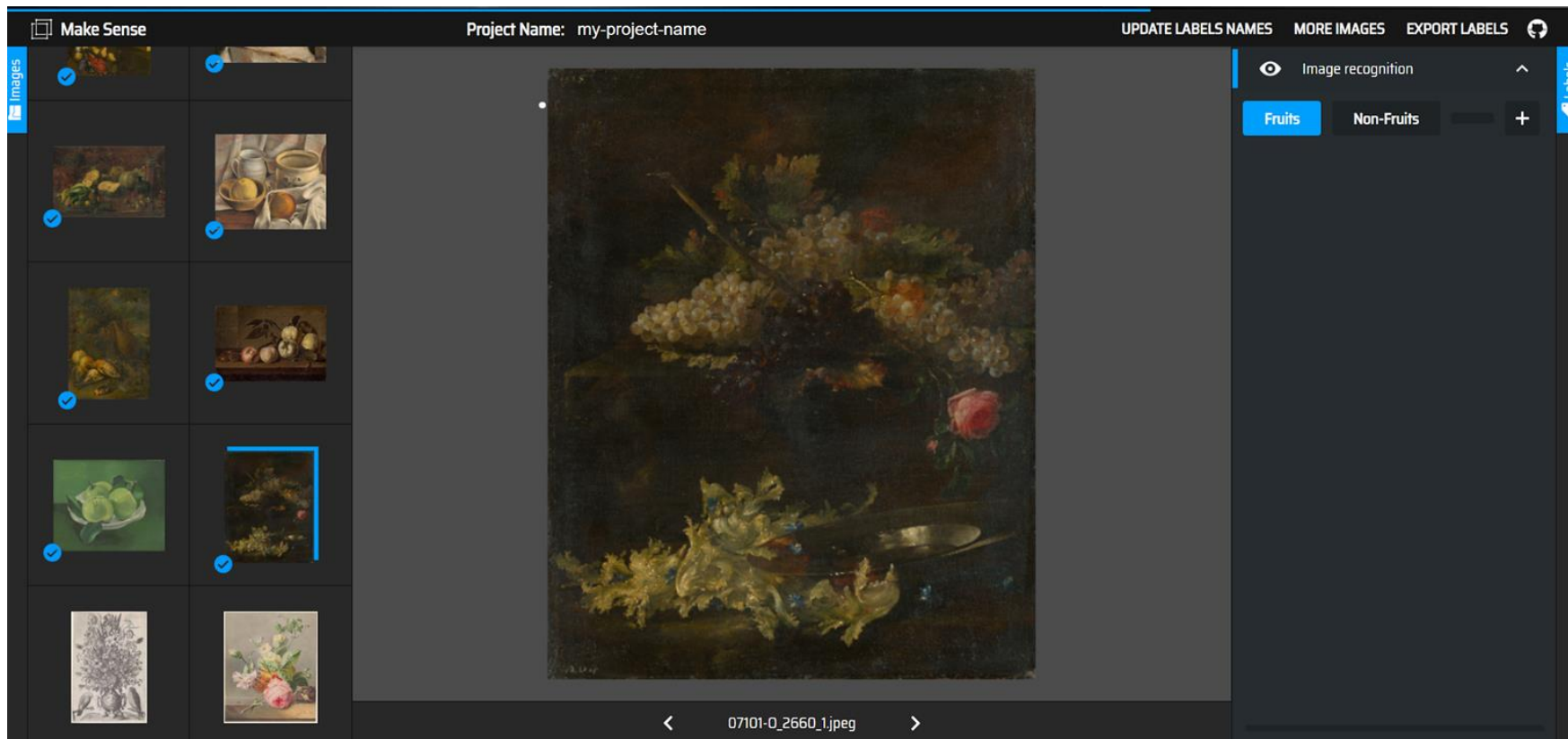
Make Sense Project Name: my-project-name UPDATE LABELS NAMES MORE IMAGES EXPORT LABELS

Images

Labels

Image recognition

Fruits Non-Fruits +



07101-0_2660_1.jpeg

Results

Lesson learned

- Identifying cultural aspects from the images is very challenging task
- A clear definition of the cultural concepts is crucial for inter rater agreement
- Cultural background of the annotators, gender and personal preference contributed to the low/random agreement

ChIA cultural food image game

ChIA Cultural Food Image Memory Game – find the historical and corresponding current food images that match!





ÖAW

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GO!DIGITAL
NEXT GENERATION



Thank you



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@yalemisew

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