

MUSTI - Multimodal Understanding of Smells in Texts and Images at MediaEval 2022

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Abstract

MUSTI aims to collect information about smell from digital text and image collections from the 17th to 20th century in a multilingual setting. More precisely, MUSTI studies the relatedness of evocation of smells (smell sources being identified, objects being detected, gestures being mentioned or recognized) between texts and images. The main task is a binary classification task and entails identifying whether a pair of image and a text snippet contains the same smell source independent of what is the smell source. An optional sub-task is the determination of the smell sources that make the respective pair related.

1. Introduction

To make sense of digital heritage collections, we need to go beyond an oculo-centric approach and to engage with their olfactory dimension as it offers a powerful and direct entry to our emotions and memories. Via the task Multimodal Understanding of Smells in Texts and Images (MUSTI) ¹ in the scope of MediaEval 2022², we aim to accelerate the understanding of olfactory references in multilingual texts and images as well as the connection between these modalities. As recent and ongoing exhibitions at Mauritshuis³ in The Hague, Museum Ulm⁴ in Ulm, and the Prado Museum⁵ in Madrid demonstrate, museums and galleries are keen to enrich museum visits with olfactory components - either for a more immersive experience or to create a more inclusive experience for differently abled visitors such as those with a visual impairment.

Reinterpreting historical scents is attracting attention from various research disciplines [1] in some cases leading to novel collaborations with perfume makers such as the Scent of the Golden Age candle developed after a recipe by Constantijn Huygens in a collaboration between historians and a perfume maker.⁶ To ensure that such enrichments are grounded in historically

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¹<https://multimediaeval.github.io/editions/2022/tasks/musti/>, retrieved on 2022/12/12.

²<https://multimediaeval.github.io/editions/2022/>, retrieved on 2022/12/12.

³<https://www.mauritshuis.nl/en/press-releases/smell-the-art-fleeting-scents-in-colour/>, retrieved on 2022/12/12.

⁴<https://museumulm.de/veranstaltung/der-nase-nach/>, retrieved on 2022/12/12.

⁵<https://www.museodelprado.es/en/whats-on/exhibition/the-essence-of-a-painting-an-olfactory-exhibition/07849a71-d9b0-faeb-94c1-689f2614f8d0>, retrieved on 2022/12/12.

⁶<https://www.huygens.fr/en/essential-oil-candles/2523-826-constantijn-huygens-candle.html>, retrieved on 2022/12/12.

correct contexts, language and computer vision technologies can aid in finding olfactory relevant examples in their collections and related sources.

While the sense of smell is of vital importance in our day-to-day lives, little attention has been paid to it within the natural language processing and computer vision communities. While there are some lexicons on smell [2, 3], the Odeuropa text benchmark dataset is the first multilingual, cross-domain text dataset focused on smell references [4]. Similarly, for computer vision, no prior datasets existed until the ODOR challenge dataset [5, 6]. The MUSTI task brings these modalities together, inviting the research community to explore parallels and complementarities in the way smells are described and depicted in different modalities. MUSTI covers English (EN), German (DE), French (FR), and Italian (IT).⁷

This report continues with a detailed task description in Section 2. The steps followed to prepare data for training and evaluation are provided in Section 3. The evaluation methodology is described in Section 4 and Section 5 concludes this report.⁸

2. Task description

The manner in which humans engage with smell is a prime example of intangible cultural heritage: the way smells are created, in what situations they are used, but also how they are appreciated are highly culturally-dependent. By engaging with expressions of smells in texts and images across multiple genres and multiple languages over a longer period of time, we can gain insights into how smells have affected human interactions through time.

Smell is an underrepresented dimension of many multimedia analysis and representation tasks. The goal of MUSTI is to further the understanding of descriptions and depictions of smells by recognizing and connecting references to smells in texts and images. In this shared task, participants are provided with multilingual texts and images, from the 17th to the 20th century, that pertain to smell (i.e., selected because they evoke smells).

Task participants develop language and image recognition technologies to predict whether a text passage and an image evoke the same smell source or not. In a subsequent optional sub-task, the participants are also asked to identify common smell source(s) between the text passages and the images. The detection of the smell source includes detecting the person, object or place that has a specific smell, or that produces odours (e.g. plant, animal, perfume, human).

The “Quest for insight” part, which goes beyond quantitative evaluation by posing questions that endorse a deeper understanding of the challenges, including data and the strengths and weaknesses of particular types of approaches, of MUSTI consists of the following questions:

1. What does it mean for a text passage and an image to be related in terms of smell?
2. Do different text and image genres reference smell differently?
3. Do different languages reference smell differently?
4. How do references to smell in texts and images change over time?
5. How do relationships between smell references in texts and images change over time?

3. Data preparation and Release

The data consists of copyright-free texts (historical books and documents) and partly copyrighted images from open repositories.⁹ We offer sentences that should be matched with images, which

⁷More information on Mediaeval 2022 and MUSTI task can be found on <https://multimediaeval.github.io/editions/2022/> and <https://multimediaeval.github.io/editions/2022/tasks/musti/> respectively.

⁸Task related materials can be found on https://github.com/Odeuropa/musti_mediaeval2022.

⁹For copyrighted materials, the source URL is shared with participants.

are selected from RKD, Bildindex der Kunst und Architektur, Museum Boijmans, Ashmolean Museum Oxford, Plateforme ouverte du patrimoine and annotated with 80+ categories of smell objects and gestures such as flowers, food, animals, sniffing and holding the nose.¹⁰

3.1. Candidate creation

In this step, we search for texts potentially related to the images used for the task. For English, the text are extracted from the British Library corpus, the Early English Books Online (EEBO), and from Project Gutenberg. For Italian, the sources are Project Gutenberg, Liber Liber, and Wikisource. For French, the sources include Gutenberg, Gallica, and the ARTFL Project. Finally, Berlin State Library OCRs and Deutsches Textarchiv are utilized for German text data.

First, each sentence in the corpora is lemmatized, then candidate extraction is based on the presence in the text of words from three different image metadata fields (when available):

- *Title*: the nouns (lemmatized) in the title, representing the subjects/objects in the painting.
- *Categories*: labels of visible smell sources, identified by the annotators within the paintings.
- *Keywords*: list of keywords associated with images in their respective collections.

We only keep sentences sharing content with at least two of the three fields. Additionally, we identified if each sentence does or does not contain a *Smell Word*, e.g. *stink*, *smell*, *odor*, *sniff*, *fetid*, *smelly*. Sentences containing a *Smell Word* are more likely to be instances where images and texts represent the same smell while sentences without *Smell Word* are usually only about the same subject without evoking any smell related to it.¹¹

3.2. Annotation methodology

We run our sentence extractor based on *categories*, *keywords*, and *titles* to create text-image pairs. Using the sentence extractor, we look for pictures containing more categories, keywords or nouns from the title to find texts more relevant as possible to the image.

For Subtask 1, we annotate matches (pairs of texts and images) as evoking the same smell (YES) or not (NO). The images have already been selected as either explicitly or implicitly evoking a smell. A YES instance is a text such as “*Having secured rooms in this establishment, we started for a walk through the town. The first thing that strikes a stranger upon his arrival at Accra is a strong, all - pervading smell of pig*” for a picture representing pigs.

The annotation NO is used when there is no match in terms of the smell represented in the text and the image, such as in “*How I do hate tobacco and that disgusting habit of smoking, cried Geraldine, who was a very fastidious little lady, and could not endure the smell of a pipe or even a cigar*” for a picture in which there is no pipes, smoke, or cigars.

The NO examples have two possible characteristics that make the matching more difficult: the *negation* cases and more *tricky* instances containing the same object(s) but not the same smell. We consider an image-text pair a negation if the object which evokes the smell represented in the picture is negated in the text. An example of negation is the text “*They had no censers to perfume the air extinguishing the morning fragrance, nor bore they their diamond crosiers through the streets*” for a picture in which there are no censers.

The tricky cases contain all the cases in which the smell-evoking object is mentioned in the text but does not evoke the same smell represented in the picture. These are metaphorical or similarity sentences like “*Will you not believe these miracles? Which however of themselves they*

¹⁰The taxonomy is described in https://odeuropa.eu/wp-content/uploads/2022/05/D2_1_TaxonomyOfOlfactoryPhenomena.pdf.

¹¹The translations of the English taxonomy in respective languages are utilized.

may shine like a candle lighted up” for a picture representing candles. Sentences mentioning the same smell source represented in the picture but with a different acceptance “*The barracks and all buildings were heaps of ruins, the fires still burning, the smoke and stench from which were offensive and suffocating*” for a picture representing a different kind of smoke such as the smoke of a pipe. Finally, sentences mentioning the same object as picture but in a different state “*The Sunday or visiting dress of Mrs. Yates consisted of a thick sort of silk [...] printed in a large chintz pattern on a white ground , in which butterflies and flowers, of unknown and fantastic varieties, predominated.*” are considered as tricky.

In Subtask 2, we add a second layer of annotations to the image-text pairs annotated with YES in the Subtask 1. In this case, the annotation indicates the person, object or place (or more than one) related to the evoked smell appearing both in the text and in the image. For instance in “*Having secured rooms in this establishment , we started for a walk through the town . The first thing that strikes a stranger upon his arrival at Accra is a strong , all - pervading smell of pig.*”, the smell evoking element appearing both in the text and in an image retrieved from RKD¹² “pig”.

3.2.1. Data release

The data is released as training and test sets, with a similar proportion of YES and NO labels. The YES instances are labeled for subtask 2 as well. The number of instances in the training and test data are 795 and 200 for EN, 480 and 213 for DE, 300 and 200 for FR, and 799 and 201 for IT.

4. Evaluation

We utilize the binary F1-macro score as provided by Scikit-learn library.¹³ for Subtask 1. A baseline F1-macro score that is based on the majority label as prediction is 27.14 on the whole test set, which is 42.85 for EN, 42.89 for DE, 22.07 for FR, and 42.73 for IT. Subtask 2 is evaluated by averaging F1-macro scores calculated for predictions of each instance. The F1-macro calculation is performed on the overlap between gold and predicted labels. The label mismatch between gold and predicted labels is resolved by creating a label set by unifying and alphabetically sorting what appears both in gold and predicted labels for each prediction.

5. Conclusion

We propose the MUSTI challenge for detecting text-image pairs that contain references to the same smell source(s) as two subtasks, i.e. prediction of the existence of the relationship as a binary classification task and detecting the smell-related tokens that are connected through this relationship. The texts and images are extracted from historical archives. Moreover, We identified tricky examples of text during the annotation phase. Finally, a baseline based on prediction as majority label was calculated for the test data released. The system performances on MUSTI dataset are reported in [7] and [8].

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¹²<https://rkd.nl/nl/explore/images/193733>, , retrieved on 2022/12/20.

¹³https://scikit-learn.org/stable/modules/generated/sklearn.metrics.f1_score.html, retrieved on 2022/12/12.

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