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# OCR17: Ground Truth and Models for 17th century French Prints (and hopefully more) 

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#### Abstract

Machine learning begins with machine teaching: in the following paper, we present the data that we have prepared to kick-start the training of reliable OCR models for 17th century prints written in French. The construction of a representative corpus is a major challenge: we need to gather documents from different decades and of different genres to cover as many sizes, weights and styles as possible. Historical prints containing glyphs and typefaces that have now disappeared, transcription is a complex act, for which we present guidelines. Finally, we provide preliminary results based on these training data and experiments to improve them.


## Keywords

OCR, 17th century French, training data, transcription guidelines, corpus building

## I INTRODUCTION

OCR engines such as Abbyy ${ }^{1}$ or Tesseract [Smith, 2007] today come with models that work perfectly for the most recent documents written in French (i.e. the 19th and 20th centuries) ${ }^{2}$, but older periods are still less well handled by the machines. Our objective is therefore to solve this problem, and to propose the tools, as much as the data and the method necessary for the processing of historical documents following the recommendations on open science [Chagué et al., 2020], and especially prints in French of the 17th century. In doing so, we hope to prepare the digitisation of the entire period of the Ancien Régime, the prints of the 18th century and the second half of the 16th century written in French being relatively similar to those of the 17th century.

The reliability of OCR models depends on both the quantity and the quality of training data. On the one hand, quantity needs to be produced and made freely available to other scholars, which is sadly not always the case. On the other hand, quality needs to be properly defined, since philological traditions vary from one place to another [Duval, 2018], but also from one period to another [Gabay, 2014, Duval, 2015]. Both problems need therefore to be addressed to propose a reliable solution for the OCRisation of historical prints written in French.

Following the example of GT4HistOCR [Springmann et al., 2018], which mainly focuses on German (and marginally Latin [Springmann et al., 2016]), we have designed a corpus of Ground

[^0]Truth (GT) made of c. 30,000 lines taken from 37 prints in French of the 17 th century (tab. 9). These documents have been carefully chosen so that they contain different kinds of typefaces (style, weight, size), and thus cover a maximum of the fonts possibly used for this type of document. Because graphetic variants that have now disappeared may have existed (such as the long $s: f$ ), transcription is a particularly technical act for historical documents, and we describe our transcription guidelines. Based on these data, we offer robust state-of-the art models for two open source OCR engines, both available to users via simple interfaces: Kraken [Kiessling, 2019]/eScriptorium [Kiessling et al., 2019] and Calamari [Wick et al., 2020]/OCR4all [Reul et al., 2019].

## II CORPUS BUILDING

Several corpora exist today for 17th century documents. We have already mentioned GT4HistOCR [Springmann et al., 2018], but others are available such as the IMPACT Dataset [Papadopoulos et al., 2013], of which $80 \%$ of the documents date however from the 19th and 20th centuries. The first corpus is mainly focusing on German ${ }^{3}$, as we previously mentioned, such as the RIDGES dataset [Springmann and Lüdeling, 2017], and the second corpus contains only about $15 \%$ of documents in French. In both cases we do not control (or know precisely) the transcription guidelines, which is an important philological problem. We have therefore decided to create our own corpus.

Producing training data in order to kick-start the creation of a generic model for 17 th c . documents written in French implies the gathering of various sources, which can be selected in many ways, from piling up data from different projects to the scrupulous association of complementary sources. For our project, due to the paucity of available data, we chose to follow the second option, which is not a simple task. We therefore had to define a method to select the documents that should be included in our corpus.

Table 1: Distribution of the prints in the training corpus per decade

| Decade | Total items | Total lines |
| :--- | :--- | :--- |
| 00's | 1 | 617 |
| 10's | 1 | 198 |
| 20's | 3 | 2,689 |
| 30's | 5 | 3,159 |
| 40's | 5 | 3,527 |
| 50 's | 3 | 2,008 |
| 60's | 5 | 5,089 |
| 70 's | 4 | 3,836 |
| 80 's | 5 | 3,336 |
| 90 's | 5 | 3,709 |

Table 2: Distribution of the prints in the training corpus per genre

| Genre | Total |
| :--- | :--- |
| Drama | 17 |
| Poetry | 4 |
| Novel | 3 |
| Letter | 2 |
| Philosophy | 2 |
| Physics | 2 |
| Sermon | 1 |
| Theology | 1 |
| Travel | 1 |
| Maxims | 1 |
| Medicine | 1 |
| Memoirs | 1 |
| Mechanics | 1 |

[^1]Since the advent of corpus pragmatics, linguists have been working on how to associate data to obtain representativeness, i.e. "the extent to which a sample includes the full range of variability in a population" [Biber, 1993], but such a notion is now more and more debated [Raineri and Debras, 2019]. Following the example of corpus linguists using extralinguistic criteria (sociological, demographic...) [Crowdy, 1993], we have decided to select samples mostly according to bibliographical metadata (printing date and place, literary genre, author...), which serve as a proxy for paleographical information - a good diachronic distribution should for instance ensure a correct representation of the very diverse typographical material (e.g. fig. 1). We also took into account digital information (size and resolution of the images), in addition, of course, to a careful philological analysis of the documents.


Figure 1: Mixing styles, heights and casing in 17th c. French prints

Prints production dates are distributed over the century, with a special attention for books printed between 1620 and 1700 (tab. 1) because it covers one of the most important periods in the literary history of France, that of classical French. Regarding genre, the result can be seen as a two-tier corpus (tab. 2), with a primary one consisting of literary texts (drama, poetry, novels...) and a secondary one made of scientific works (medicine, mechanics, physics...). If the vast majority has been printed in Paris, we have also included books coming from Belgium (Brussels) and Holland (Leiden), which were major production centres at the time [van Eeghen, 1960-1978, Eisenstein, 1992] ${ }^{4}$.


Figure 2: Tristan L'Hermite, Panthée, 1639

|  | Lower | Upper | Total |
| :---: | :---: | :---: | :---: |
| Dramatic | 396,984 | 43,295 | 440,279 |
| texts | $90.17 \%$ | $9.83 \%$ |  |
| Non-dramatic | 297,527 | 12,544 | 310,071 |
| Texts | $95.96 \%$ | $4.04 \%$ |  |

Table 3: Percentage of uppercase letters in dramatic texts vs. non-dramatic texts in our dataset.

As we can see, the corpus is not balanced, since not only literary texts, but also plays are clearly

[^2]over-represented. Such a choice has been made for two reasons. On the one hand, we need GT in italics, and since versified texts use this type of style a lot [Speyer, 2019], we increase the amount of data in italics by selecting plays in verse. On the other hand, we must also add capital letters in the GT, and therefore find examples of this type of character: plays, once again, are an abundant source with the names of the speakers written in capitals (fig. 2 and tab. 3). Such a strategy should help us deal with highly complex layouts (fig. 1).

Regarding the resolution, images used can be divided into three classes: 72 (20 prints), 400 (14 prints) and 600 dpi ( 1 print) (fig. 3). Indeed, many scans available online are in low resolutions (usually 72 dpi , an older computer standard introduced by Apple in the 1980s), which introduces significant changes in the shape of letters (fig. 4) that our model needs to handle properly.


Figure 3: Examples of GT with different resolutions

Figure 4: Impact of the resolution on the letter $e$, which can be confused with $c$ because of the disappearance of the eye.

## III TRANSCRIPTION GUIDELINES

Theoretical background. Transcription is a very delicate matter: more than copying, transcribing has to be understood as an act of translation [Robinson and Solopova, 1993] and, as the saying goes, traduttore, traditore. Following Robinson and Solopova, there are four different levels of transcription, rearranged into two categories by D. Stutzmann [Stutzmann, 2011]: those that describe the image (called "graphic" and "allographetic" transcriptions) and those that describe the text (called "graphemic" and "regularised"). For our project, we exclude the two most extreme types of the spectrum: graphic transcription (which retains all the visual richness of the original) would be far too time-consuming, and regularised transcription (which fully aligns the spelling to a standard) would be linguistically too poor. Only the allographetic and the graphemic transcriptions will therefore interest us.

The allographetic transcription aims at keeping all the graphetic richness: it reduces all the graphic variants to extended types, and thus gives access to various forms of each letter or sign. For instance, the distinction between $\langle\uparrow\rangle$ and $\langle\mathrm{s}\rangle$ is kept because they are two graphetic variants of $s$, but not the distinction between $\langle\downarrow$ with or without a leftward swash, because they both would be graphic variations of the $\langle\uparrow\rangle$. Long impossible for material reasons, such transcriptions are now accessible to all researchers thanks to projects such as the Medieval Unicode Font Initiative (MUFI, Haugen [2015]), that have played an important role by designing and pushing new

Unicode code points to the Unicode standard [Consortium, 2019] ${ }^{5}$.
The graphemic transcription further limits variation by reducing the different possible types to their meaning in the alphabetic system. Unlike the allographetic approach, graphemic transcription is better known to philologists because it resembles the traditional semi-diplomatic transcription. Scholars have clarified its execution framework, and have been using it in their editions for a long time. In practice, words like eftoit are transposed as estoit in the edited version, but not as était which would be a regularised transcription. A few points of detail remain however debated, in particular concerning the need to expand the abbreviations when adopting a graphemic approach ${ }^{6}$.
Project framework. As our intent for our data is to produce OCR models and to transcribe automatically print to do more research on the French language used in these prints, our transcription guidelines lies on the side of the graphemic transcription, without regularisation. However, we introduced few graphetic concerns, listed below, because graphetic variations can be a linguistic evidence, such as the long $s$, which inform us of the content of the type boxes used by printers ${ }^{7}$. The result is a hypbrid graphemic-allographetic transcription with very punctual regularisations.

```
fon interprete & Protecole en fes Ef.
cripts; ( s'il n'eft làmefme celuy de So-
crates, fon plus diuin Precepteur) leur
```

> fon interprete \& Protecole en fes Efcripts; (s'il n'eft là mefme celuy de Socrates, fon plus diuin Precepteur) leur

Figure 5: Excerpt and transcription example of Marie de Gournay, Egalité, 1622

Normalisation of spelling. Our choice leads us to keep the original spelling of the source (e.g. fig. 5). We include in spelling the absence of normalisation for letters such as $\langle\mathrm{u}\rangle\langle\langle\mathrm{v}\rangle$ and <i>/<j>, whose usage was different from the current one (no consonant/vowel distinction): e.g., we transcribe diuin where one would normalise it as divin. We respect accent absence (e.g. interprete and not interprète), but we transcribe dotless i ( $1>$ ) as 〈i》, as it is in many cases a printing problem. To avoid confusion for the machine, commas used as cedillas (FRANC,OIS and not FRANÇOIS) and apostrophes used as accents (ARME'ES and not ARMÉES) are kept as they are, and not regularised as accents or cedillas. Historical spellings (e.g. Efcripts, normalised Ecrits) and calligraphic letters (e.g. celuy, normalised celui) are kept.
Variation of letters and ligatures. As mentioned, we keep one allographetic variation: the long $s$ (e.g. mefme and not mesme). Other variations are ignored. Aesthetic ligatures that still exist in French (e.g. «œ» vs «oe») have been encoded, but not those that have disappeared despite their possible existence in Unicode (e.g. $\langle\mathfrak{f}\rangle)^{8}$. Examples are provided in Table $4^{9}$.

[^3]About spacing．Spacing is a problem because the compositor can＂pack＂the words so that they all fit into the line space．It is therefore typographical information that must be treated with care，but in most cases we follow the graphemic approach，which tends to distinguish units grammatically rather than graphically，while retaining period peculiarities．

## IV EXPERIMENTS

General set－up．In order to train and evaluate models，we use a regular $80 \%$ of the produced dataset for training， $10 \%$ for development purposes and $10 \%$ for evaluation．The split is produced at the level of each print，resulting de facto as a in－domain test．

| Category | Description | Status | Transcription | Example |
| :---: | :---: | :---: | :---: | :---: |
| Ligature | Ligature $\mathrm{O}+\mathrm{E}$ «œ＞ | Graphetic | U＋0153／U＋0152 | cour |
| Ligature | Ligature $\mathrm{A}+\mathrm{E}$＜æ＞ | Graphetic | U＋00E6／U＋00C6 | Cx－Egyfte |
| Ligature | Ligature long S＋T＜ ft ＞ | Graphemic | No ligature | Chreftienté． |
| Ligature | Ligature L＋L＜ll＞ | Graphemic | No ligature | eternelle |
| Ligature | Ligature $\mathrm{C}+\mathrm{T}$＜ct＞ | Graphemic | No ligature | Edict： |
| Ligature | Ligature $\mathrm{S}+\mathrm{P}$＜sp＞ | Graphemic | No ligature | l＇esprit |
| Ligature | Ligature long S＋L＜fl＞ | Graphemic | No ligature | meflent |
| Ligature | Ligature U＋S＜us＞ | Graphemic | No ligature | indiuidus |
| Ligature | Ligature S＋I＜fi〉 | Graphemic | No ligature | quafi |
| Ligature | Ligature long S＋long S $\langle f$ ¢ | Graphemic | No ligature | ieuneffe |
| Ligature | Ligature $\mathrm{F}+\mathrm{F}+\mathrm{I}$ 〈ffi〉 | Graphemic | No ligature | fuffifance |
| Ligature | Ligature $\mathrm{I}+\mathrm{S}$＜is» | Graphemic | No ligature | puis |
| Allograph | Capital E | Graphemic | U＋0045 | $E t E t$ |
| Allograph | Capital A | Graphemic | U＋0041 | A eAu |
| Allograph | Capital M | Graphemic | U＋004D | croy Mais chas |
| Allograph | Small E with long finial | Graphemic | U＋0065 | espece |
| Allograph | Tittle as tilde or dot | Regularised |  | je jugé |
| Allograph | Small long and short S | Graphetic | U＋017F | fans |
| Abbreviation | Combining tilde＜） | Graphetic | U＋0303 | d＇autát deuós |
| Abbreviation | Combining Macron＜${ }^{\text {¢ }}$ | Graphemic | U＋0303 | väteric |
| Abbreviation | Ampersand＜\＆〉 | Graphetic | U＋0026 | 8 |
| Diacritics | Combining vertical line \ll | Regularised |  | $\dot{\text { a }}$ |
| Diacritics | Apostrophe | Graphetic | U＋0027 | ARMEES |
| Diacritics | Comma | Graphetic | U＋002C | FRANC，OIs |
| Hyphenation | Hyphen | Codified＜${ }^{\text {¢ }}$＞ | U＋00AC | foigneu timbecile，di－ |

Table 4：Main transcription choices

We additionally produced 4 others small samples for out-of-domain testing based on centuries, from the $16^{\text {th }}$ c. to the $19^{\text {th }}$ (cf. tab. 10,11, 12 for details, tab. 5 otherwise). We specifically designed these out-of-domain samples to exclude gothic and other special fonts such as civilités ones (cf. fig. 6), as our training corpora only include roman or italic typefaces.

| Dataset | Caracters |
| :---: | ---: |
| 17th c. | 91,104 |
| 16th c. | 18,542 |
| 18th c. | 16,691 |
| 19th c. | 13,103 |

Table 5: Description of test sets, character counts are in NFC.
foufiours riant, fouftours Bentity'ate fant aso cbafcun, touftours fe guabes fant, foufioute diffimufat fon dinin $\mathrm{fcas}_{s}$
(a) Rabelais, Garguantua, 1535, gothic typeface

(b) Trissino, Sophonisba, 1559, civilité typeface

Figure 6: Non-selected typfaces
Two separate open-source OCR engines are used for training OCR models, namely Kraken [Kiessling, 2019] and Calamari [Wick et al., 2020]. Both tools were used in order to leverage their various differences in order to produce the best model possible. Default engine model architecture as well as hyperparameters were used for the baseline model.

Kraken Experiment: artificial lines vs. synthetic data. Texts were normalised using unicode's decomposition normalisation (NFD). This results in splitting characters such as «é> into two characters $\langle\mathrm{e}\rangle+\langle\overline{\text { ¢ }}$ (combining acute accent, U+0301). This has become in the French DH community of Kraken the de facto choice for French language OCR.

To improve the efficiency of the engine, two additional experiments have made. On the one hand, we tested a larger model architecture than the base one ${ }^{10}$, doubling the filter size of each convolutional layer, respectively from 32 to 64 and from 64 to 128 , to handle the heterogeneity of the training data. On the other hand, we used a synthetic training set on top of the manually compiled one with 27 different fonts ${ }^{11}$.
Calamari Experiment: Multiple voters and data augmentation. Regarding Calamari, we have tested another type of unicode normalisation (NFC) making sure that diacritics are combined ${ }^{12}$. We replicated here the successful protocol from Reul et al. [2018] by combining model fine-tuning (FT) - i.e. building from existing models (historical non-French Antiqua) instead of starting the training from scratch -, voting (VT) - i.e. training five models instead of one and combining their outputs during predictions -, and data augmentation (DA), - i.e. generating modified images of the input lines by blurring them, stretching them, etc.

[^4]
## V ANALYSIS

| Augm. Architect. | Artif. <br> Data | 17th c. | 16th c. | 18th c. | 19th c. | Pretrain. | Voters | Data Augm. | 17th c. | 16th c. | 18th c. | 19th c. |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| - | - | 97.47\% | 97.74\% | 97.78\% | 94.50\% | - | 1 | - | 98.47\% | 98.14\% | 98.27\% | 93.11\% |
| Yes | - | 97.92\% | 98.06\% | 97.78\% | 94.23\% | Yes | 1 | Yes | 98.76\% | 98.49\% | 96.47\% | 97.05\% |
| - | Yes | 96.65\% | 97.26\% | 97.74\% | 95.50\% | Yes | 5 | Yes | 99.05\% | 98.68\% | 98.78\% | 97.05\% |
| Yes | Yes | 97.26\% | 97.68\% | 97.84\% | 94.84\% |  |  |  |  |  |  |  |

Table 6: Accuracy (1-CER) for the experiment with Kraken.

Table 7: Accuracy (1-CER) for the experiment with Calamari

Considering the (deliberately) extreme heterogeneity of our data, such scores are promising (cf. tab. 6 \&7). However, it is clear that, regarding Kraken, synthetic data did not improve results at all, except for 19th c., and might actually in some cases lowered the score (specifically for the in-domain test). Kraken however benefited from a larger model, and this change impacted also out of-domain results except from later one (18th and 19th c.). Calamari shows again that the protocol from Reul et al. [2018] is beneficial to the results and incremental (multiple voters enhance the results of the already better ones from data augmentation and pretraining).

Despite being focused on the 17th c., the dataset is able to produce model resistant to changes in neighbouring centuries. We see that in both case, the accuracy drops by less than one percentage point. This is definitely due to the filtering of gothic fonts and special typefaces of the 16th c . prints, but also to the limited changes in common typefaces between these centuries. As for 19th century, the score dropped more for Kraken ( -2 to -3 percentage points) than for Calamari ( -1.7 points) for their best performing models on other centuries. Only the use of artificial data allowed for performance gains on 19th c. for Kraken, most probably due to the regularity provided by them.

Confusion table from Calamari (cf. tab. 14, 13, 15, 16, 17) shows an important issue with spacing recognition, and as such, word segmentation. This could be linked to both the density of the composition in early modern prints or because of the instability of the graphic segmentation, some words being sometimes welded (puisque) and sometimes not (puis que, cf. tab. 8).

| Spacing error | Example | Transcription |
| ---: | ---: | :--- |
| Composition | lemét, quilofepretendre, | lemẽnt, qu'il ofe [NO SPACE] pretendre |
| Graphic segmentation | letrióphe; mais puis quece | le triõphe; mais puis [SPACE] que ce |

Table 8: Possible sources of word segmentation errors. The token SPACE indicates the problematic zone.

Another important source of error is linked to the $\langle\mathfrak{f}\rangle$, which, once again, can be linked to paleographic problems (confusion $\langle\mathfrak{f}\rangle\langle\mathrm{f}\rangle$ ). The confusion $\langle\mathfrak{f}\rangle\langle\mathrm{s}\rangle$ might be related to the language model overtaking the OCR, or more simply, input errors on the side of the GT ${ }^{13}$.

Both issues can be treated with post-processing steps. Segmentation or $\langle\mathrm{f}\rangle /\langle\mathrm{s}\rangle$ confusion can be approached as a character classification from a pure natural language processing point of view, as shown per Clérice [2020]. In this paper, content were encoded at the character level with a percharacter binary classification (word boundary vs. in-word content) which resulted in very high

[^5]accuracy. The same process could be applied to both type of confusions. However, regarding the $\langle\mathrm{f}\rangle /\langle\mathrm{s}\rangle$ confusion, another option would be to drop the differentiation of the allographs at training and testing time, until enough GT has been produced to avoid this kind of issues.

## VI FUTURE WORK

The diachronic efficiency of the model can be improved by adding data for more recent prints: c. 20,000 additional lines will be added, to carry further tests on the creation of a model for French prints in general, and not only modern prints. Out-of-domain tests sets composed of non-francophone prints should also be created to test the efficiency of the model on similar prints in other languages.
While creating the GT, we have corrected the layout of each image. Alto and PageXML will be used to train a segmenter, the importance of which must not be underestimated since it is on its result that the OCR is performed.

## AUTHOR CONTRIBUTIONS

S.G. designed the research project, built the corpus and prepared the data for training. T.C. helped all along the process providing advice, scripts and feedback. C. R. performed the experiments with Calamari. All authors discussed the results and contributed to the final manuscript.

## DATA

Training data is available online (10.5281/zenodo.3826894): it contains all the GT used to train models, and it is distributed with a CC-BY licence. Ongoing research on OCR, with additional data and scripts, is available on Github (https://github.com/e-ditiones/OCR17).

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## ADDENDUM

This article has been originally written and submitted in 2020, and it was corrected in 2023. While we received reviews early, we were not able to complete the correction proposed in due time. We still think that this paper paints an interesting state of OCR and HTR at the time.
However, since then, tools have evolved, and the dataset has evolved has well. It became the OCR17+ dataset ${ }^{14}$, using ALTO-XML representation instead of line based segmentation. It has been largely completed since then through project such as GalliCorpora ${ }^{15}$. Kraken has since then adopted augmentation of images, and uses larger line input rather than higher convolution filters to reach better results.

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| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Ellain | Advis sur la peste | 1606 | Paris | D. Douceur |  | 617 | No | 400 | $1496 \times 2560$ | BNF | cb303981499 |
| Regnier | Les Satyres | 1612 | Paris | Toussaint Du Bray |  | 198 | Yes | 400 | $1562 \times 2580$ | BNF | cb31189430j |
| Gournay | Egalité des hommes et | 1622 |  |  |  | 824 | No | 400 | $1666 \times 2634$ | BNF | cb30529274x |
| de Viau | ©uvres | 1623 | Paris | J. Quesnel |  | 851 | No | 400 | $1334 \times 2600$ | BNF | cb34804166g |
| Balzac | Lettres | 1624 | Paris | Toussaint Du Bray |  | 1014 | No | 72 | $4267 \times 6667$ | BNF | cb300515241 |
| Descartes | Discours de la méthode | 1637 | Leiden |  | J. Maire | 431 | No | 72 | $2479 \times 3508$ | BNF | cb30328384x |
| Scudéry | L'Amour tirannique | 1639 | Paris | A. Courbé | M. Brunet + J. de La Coste | 860 | No | 72 | $4267 \times 5513$ | BNF | cb31341723p |
| L'Hermite | La Mariane | 1639 | Paris | A. Courbé | M. Brunet | 673 | No | 72 | $3796 \times 5860$ | BNF | cb39333461s |
| L'Hermite | Panthée | 1639 | Paris | A. Courbé | M. Brunet | 897 | No | 72 | $4267 \times 6100$ | BNF | cb314972698 |
| Rotrou | La Belle Alphrede | 1639 | Paris | A. de Sommaville | A. Coulon | 298 | Yes | 400 | $2570 \times 3695$ | BNF | cb31251853x |
| Scudéry | Ibrahim | 1641 | Paris | A. de Sommaville |  | 1671 | No | 72 | $2479 \times 3508$ | BNF | cb31341849n |
| Fr. de Sales | Introduction à la vie | 1641 | Paris | Imprimerie royale |  | 617 | No | 400 | $4213 \times 6084$ | BNF | cb30460001n |
| Scarron | Typhon | 1644 | Paris | T. Quinet |  | 196 | Yes | 400 | $2746 \times 3608$ | BNF | cb31308401d |
| Scarron | Le Jodelet | 1645 | Paris | T. Quinet | A. Coulon | 268 | Yes | 400 | $265 \times \times 3424$ | BNF | cb31308475z |
| Pascal | Expériences nouvelles | 1647 | Paris | P. Margat |  | 775 | No | 400 | $1684 \times 2637$ | BNF | cb31062878c |
| Voiture | Oeuvres | 1650 | Paris | A. Courbé |  | 359 | Yes | 400 | $2794 \times 3729$ | BNF | cb31600370j |
| Scudéry | Clélie | 1656 | Paris | A. Courbé |  | 897 | No | 72 | $2479 \times 3508$ | BNF | cb31341819q |
| Chapelain | La Pucelle | 1656 | Paris | A. Courbé |  | 752 | No | 400 | $4504 \times 6589$ | BNF | cb365764947 |
| Pascal | Traictez de l'équilibre | 1663 | Paris | G. Desprez |  | 972 | No | 600 | 2083×3634 | BNF | cb31081848m |
| Molière | L'Escole des femmes | 1663 | Paris | L. Billaine | J. Hénault +Cl . Blageart | 1074 | No | 72 | $4058 \times 6923$ | BNF | cb30958651f |
| Bussy-Rabutin | Histoire amoureuse | 1665 | Bruxelles | Fr. Foppens |  | 876 | No | 72 | $4267 \times 7542$ | BNF | cb36117831r |
| Molière | George Dandin | 1669 | Paris | J. Ribou | Cl. Audinet | 1323 | No | 72 | $4042 \times 7200$ | BNF | cb30958651f |
| Racine | Les Plaideurs | 1669 | Paris | Cl. Barbin | Cl. Blageart | 894 | No | 72 | $4109 \times 7643$ | BNF | cb311693885 |
| Racine | Oeuvres, t. 2 | 1676 | Paris | J. Ribou | J.-B. (I) Coignard | 1309 | No | 72 | $4267 \times 7783$ | BNF | cb31168676r |
| Racine | Oeuvres, t. 1 | 1676 | Paris | J. Ribou | J.-B. (I) Coignard | 561 | No | 72 | $4267 \times 7821$ | BNF | cb31168676r |
| La Fayette | Princesse de Clèves | 1678 | Paris | Cl. Barbin |  | 948 | No | 72 | $4267 \times 7186$ | BNF | cb307135973 |
| Racine | Oeuvres, t. 1 | 1679 | Paris | D. Thierry | D. Thierry | 1810 | No | 72 | $3767 \times 6583$ | BSB | BV012474970 |
| Pradon | Statira | 1680 | Paris | J. Ribou | Cl. Blageart | 1053 | No | 72 | $4085 \times 6956$ | BNF | cb311463583 |
| Papin | La Manière d'amolir | 1682 | Paris | E. Michallet |  | 547 | No | 400 | $1468 \times 2426$ | BNF | cb31056545b |
| Bossuet | Oraison funebre | 1683 | Paris | S. Mabre-Cramoisy |  | 769 | No | 400 | $3320 \times 4584$ | BNF | cb36575655n |
| Donneau de Vizé | Voyage des ambassadeurs | 1686 | Paris | Au Palais |  | 161 | Yes | 415 | $1500 \times 2416$ | BNF | cb303484582 |
| La Bruyère | Caractères | 1688 | Paris | E. Michallet |  | 806 | No | 72 | $4267 \times 7258$ | BNF | cb31452154x |
| Molière | Dom Garcie de Navarre | 1694 | Bruxelles | G. de Backer |  | 723 | No | 72 | $1006 \times 1768$ | ÖNB | AC10132063 |
| Boyer | Méduse | 1697 | Paris | Académie de mus. | C. Ballard | 886 | No | 72 | $3854 \times 5485$ | BNF | cb30152139c |
| Pradon | Oeuvres | 1697 | Paris | Th. Guillain | Ch. Journel | 932 | No | 72 | $4080 \times 6924$ | BNF | cb38652730w |
| Racine | Oeuvres, t. 1 | 1697 | Paris | D. Thierry | D. Thierry | 1046 | No | 72 | $2457 \times 2149$ | BNW | 7805546 |
| Bussy-Rabutin | Mémoires, t. 1 | 1698 | Paris | J. Anisson |  | 122 | Yes | 400 | $3128 \times 4036$ | BNF | cb393648983 |


| Author | Title | Date | Place | Publisher | Lines | LigaturesDPI |  | Size | Library | ID |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Bartas | La Sepmaine | 1578 | Paris | M. Gadoulleau | 62 | No | 415 | 2840×3880 | BNF | cb303572930 |
| Beroalde | Avantures de Floride | 1594 | Tours | J. Mettayer | 63 | No | 400 | $1174 \times 2186$ | BNF | cb30092726b |
| Calvin | Institution de la religion | 1562 | Geneve | G. Bourgeois | 54 | No | 400 | $2633 \times 4078$ | BNF | cb365761545 |
| Du Bellay | La Defence et illustration | 1549 | Paris | A. 1'Angelier | 58 | No | 400 | $1589 \times 2445$ | BNF | cb11968311h |
| Du Fail | Discours d'Eutrapel | 1585 | Rennes | N. Glamet | 65 | No | 416 | $1596 \times 2576$ | BNF | cb30367435k |
| Rabelais | Tiers Livre | 1546 | Paris | Ch. Wechel | 46 | No | 400 | $1573 \times 2647$ | BNF | cb31167405f |
| Ronsard | Les Amours | 1552 | Paris | Vve de la Porte | 59 | No | 400 | $1678 \times 2711$ | BNF | cb432409623 |
| Table 10: Testing data, 16 th c . |  |  |  |  |  |  |  |  |  |  |
| Author | Title | Date | Place | Publisher | Lines | LigaturesDPI |  | Size | Library | ID |
| Buffon | Histoire naturelle | 1750 | La Haye | P. De Hondt | 112 | No | 400 | 3756×4582 | BNF | cb301741874 |
| Laclos | De la Monarchie | 1791 | Paris | Impr. nationale | 79 | No | 400 | $2274 \times 3430$ | BNF | cb302389989 |
| Diderot | Essais sur la peinture | 1785 | Paris | Fr. Buisson | 56 | No | 400 | 1881×2903 | BNF | cb44312299p |
| Marivaux | Le jeu de l'amour et du hazard | 1730 | Paris | Briasson | 59 | No | 400 | $1440 \times 2609$ | BNF | cb30886471g |
| Montesquieu | Lettres persanes | 1721 | Amsterdam |  | 48 | No | 400 | $1452 \times 2588$ | BNF | cb1 19437548 |
| Rousseau | Les Pensées | 1764 | Amsterdam |  | 60 | No | 400 | $1664 \times 2904$ | BNF | cb31257216h |
| Voltaire | Zadig | 1748 | Paris/Nancy | L.-Fr Prault/A. Leseure | 44 | No | 600 | $2022 \times 3676$ | BNF | cb316044160 |

Table 11: Testing data, 18th c.

| Author | Title | Date | Place | Publisher | Lines | LigaturesDPI |  | Size | Library | ID |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Chateaubriand | Atala | 1801 | Paris | Migneret/Dupont | 41 | No | 600 | $1914 \times 3280$ | BNF | cb30227639h |
| Constant | Adolphe | 1816 | Paris/Londres | Treuttel et Würtz/H. Colburn | 40 | No | 400 | $1560 \times 2653$ | BNF | cb319643212 |
| Flaubert | Salammbô | 1863 | Geneve | M. Lévy frères | 61 | No | 400 | $2264 \times 3348$ | BNF | cb304403988 |
| Gautier | Le Roman de la momie | 1858 | Paris | L. Hachette | 57 | No | 400 | $1605 \times 2739$ | BNF | cb30490246s |
| Hugo | Odes | 1823 | Paris | Persan/Pélicier | 42 | No | 600 | $2208 \times 3656$ | BNF | cb32263200h |
| Musset | A quoi rêvent les jeunes filles | 1833 | Paris | E. Renduel | 59 | No | 400 | $1858 \times 3193$ | BNF | cb30999539c |
| Nerval | Scènes de la vie orientale | 1848 | Paris | F. Sartorius | 53 | No | 400 | $2300 \times 3643$ | BNF | cb32482331k |

Table 12: Testing data, 19th c .

| GT | PRED | COUNT | PERCENT |
| ---: | :---: | ---: | ---: |
| $\}$ | $\}$ | 39 | $4.36 \%$ |
| $\{$ é $\}$ | $\{\mathrm{e}\}$ | 22 | $2.46 \%$ |
| $\{\}$, | $\{\}$. | 22 | $2.46 \%$ |
| $\{1\}$ | $\}$ | 21 | $2.35 \%$ |
| $\{$ f\} | $\{\}$ | 20 | $2.23 \%$ |
| $\}$ | $\}$ | 19 | $2.12 \%$ |
| $\{$ i\} | $\}$ | 14 | $1.56 \%$ |
| $\{$ \{\} $\}$ | $\}$ | 13 | $1.45 \%$ |
| $\}\}$ | $\}$ | 12 | $1.34 \%$ |
| $\{\mathrm{c}\}$ | $\{\mathrm{e}\}$ | 10 | $1.12 \%$ |

Table 13: Confusion table for the best Calamari models, indomain test, 17th c. prints

| GT | PRED | COUNT | PERCENT |
| :---: | :---: | ---: | ---: |
| $\}$ | $\}$ | 51 | $13.18 \%$ |
| $\{\mathrm{t}\}$ | $\{1\}$ | 32 | $8.27 \%$ |
| $\{\mathrm{f}\}$ | $\{\mathrm{f}\}$ | 21 | $5.43 \%$ |
| $\{>\}$ | $\{\mathrm{n}\}$ | 19 | $4.91 \%$ |
| $\{\mathrm{e}\}$ | $\{\mathrm{e}\}$ | 18 | $4.65 \%$ |
| $\{\mathrm{e}\}$ | $\{\mathrm{e}\}$ | 13 | $3.36 \%$ |
| $\{\mathrm{e}\}$ | $\{\mathrm{o}\}$ | 12 | $3.10 \%$ |
| $\{-\}$ | $\}$ | 11 | $5.68 \%$ |
| $\{\mathrm{c}\}$ | $\{\mathrm{e}\}$ | 8 | $2.07 \%$ |
| $\{»\}$ | $\}$ | 7 | $3.62 \%$ |

Table 16: Confusion table for the best Calamari models, out-of-domain test, 19th c. prints

| GT | PRED | COUNT | PERCENT |
| :---: | :---: | ---: | ---: |
| $\}$ | $\}$ | 33 | $12.69 \%$ |
| $\{’\}$ | $\}$ | 14 | $5.38 \%$ |
| $\{\mathrm{f}\}$ | $\{\mathrm{f}\}$ | 11 | $4.23 \%$ |
| $\}$ | $\}$ | 11 | $4.23 \%$ |
| $\{$ é $\}$ | $\{\mathrm{e}\}$ | 5 | $1.92 \%$ |
| $\{\mathrm{~s}\}$ | $\{\mathrm{f}\}$ | 5 | $1.92 \%$ |
| $\{\}$, | $\}$ | 5 | $1.92 \%$ |
| $\{\mathrm{e}\}$ | $\{$ é $\}$ | 5 | $1.92 \%$ |
| $\{\}$. | $\}$ | 4 | $1.54 \%$ |
| $\{\mathrm{a}\}$ | $\{\mathrm{a}\}$ | 3 | $1.15 \%$ |

Table 14: Confusion table for the best Calamari models, out-of-domain test, 16th c. prints

| GT | PRED | COUNT | PERCENT |
| ---: | :---: | ---: | ---: |
| $\}$ | $\}$ | 60 | $25.75 \%$ |
| $\{\mathrm{~s}\}$ | $\{\mathrm{f}\}$ | 14 | $6.01 \%$ |
| $\{\mathrm{e}\}$ | $\{\mathrm{e}\}$ | 9 | $3.86 \%$ |
| $\{\mathrm{e}\}$ | $\{\hat{e}\}$ | 7 | $3.00 \%$ |
| $\{\mathrm{u}\}$ | $\{\mathrm{n}\}$ | 4 | $1.72 \%$ |
| $\{\mathrm{c}\}$ | $\{\mathrm{e}\}$ | 4 | $1.72 \%$ |
| $\{\mathrm{e}\}$ | $\{\hat{e}\}$ | 4 | $1.72 \%$ |
| $\{\mathrm{f}\}$ | $\{\mathrm{C}\}$ | 4 | $1.72 \%$ |
| $\{-\}$ | $\{\neg\}$ | 3 | $1.29 \%$ |
| $\{\mathrm{a}\}$ | $\{\mathrm{a}\}$ | 3 | $1.29 \%$ |

Table 15: Confusion table for the best Calamari models, out-of-domain test, 18th c. prints

| GT | PRED | COUNT | PERCENT |
| :---: | :---: | ---: | ---: |
| $\}$ | $\}$ | 144 | $16.36 \%$ |
| $\{$ f $\}$ | $\{f\}$ | 36 | $4.09 \%$ |
| $\{\mathrm{t}\}$ | $\{1\}$ | 33 | $3.75 \%$ |
| $\{\mathrm{e}\}$ | $\{$ é $\}$ | 25 | $2.84 \%$ |
| $\{\mathrm{e}\}$ | $\{\mathrm{e}\}$ | 23 | $2.61 \%$ |
| $\{»\}$ | $\{\mathrm{n}\}$ | 20 | $2.27 \%$ |
| $\{\mathrm{~s}\}$ | $\{\mathrm{f}\}$ | 19 | $2.16 \%$ |
| $\}$ | $\}$ | 18 | $2.05 \%$ |
| $\{’\}$ | $\}$ | 14 | $1.59 \%$ |
| $\{\mathrm{e}\}$ | $\{\mathrm{o}\}$ | 12 | $1.36 \%$ |

Table 17: Confusion table for the best Calamari models, out-of-domain test, 16th c., 18th c., 19th c. prints


[^0]:    ${ }^{1}$ https://pdf.abbyy.com
    ${ }^{2}$ Several models are available at this address: https://github.com/tesseract-ocr/tessdata_best.

[^1]:    ${ }^{3}$ It is important to note that a large part of the German-written corpora are not only printed in antiqua but in fraktur, which considerably minimises their interest for the OCRisation of documents in French.

[^2]:    ${ }^{4} \mathrm{~A}$ detailed list of the contents of the corpus can be found in the appendix (cf. tab. 9).

[^3]:    ${ }^{5}$ In certain cases, characters which were not accepted (yet) by the Unicode governing bodies might be stored in the private zone of Unicode, being only supporter with MUFI-related font as a consequence.
    ${ }^{6}$ The Conseils pour l'édition des textes médiévaux (which are a reference for rigorous philological editing of texts, medieval or not) suggest the expansion of abbreviations [Bourgain and Vieillard, 2001, p.61] and the absence of expansion in graphemic transcriptions is presented as a "hybrid" practice by D. Stutzmann [Stutzmann, 2011, p. 251]. However, HTR data production and edition should be seen as two different tasks, and moreover, transcription and abbreviation resolution should be seen as two different computational tasks.
    ${ }^{7}$ The use of ligatures, often involving a long $s$, has slowed down the use of accents [Biedermann-Pasques, 1992, p. 92].
    ${ }^{8}$ Not all ligatures are present in the unicode standard or in MUFI: the task would therefore have been too complicated for a very limited interest.
    ${ }^{9}$ In the dataset, some folders are named with mufi: they include a richer use of unicode character. These folders weren't use for training purposes but were used to evaluate the weight of a wider transcription of allographs.

[^4]:    ${ }^{10}$ Base VGSL architecture of Kraken recognition model: [1,48,0,1 Cr3,3,32 Do0.1,2 Mp2, $2 \mathrm{Cr} 3,3,64$ Do0.1,2 Mp2,2 S1(1x0)1,3 Lbx100 Do].
    ${ }^{11}$ Namely: IM FELL English SC, IM FELL English, IM FELL Great Primer, IM FELL Double Pica, IM FELL Double Pica SC, IM FELL DW Pica, 1592 GLC Garamond, 1689GLCGaramondW00SC-Norm, Garamond, EB Garamond, EB Garamond 12 All SC, 1689 Almanach, Fournier MT Std, Bodoni 72 Oldstyle, Didot, Chapbook, DTLElzevirS, DTL Elzevir, P22 Operina Romano, Hultog, JSL Ancient, Old Claude LP Std, Chapbook, 1756DutchW01-Normal, 1726RealEspanolaW01-Rg, 1776_Independence, Palatino.
    ${ }^{12}$ As a result, score between both engines are not comparable, as they do not use the same unicode normalisation which results in a different number of evaluated characters.

[^5]:    ${ }^{13}$ As a reviewer kindly said, "muscle memory" is sometimes quite strong.

[^6]:    14https://github.com/Heresta/OCR17plus.
    ${ }^{15}$ https://github.com/Gallicorpora

