





# Pattern Recognition

## "Final"

## **Team No. 14:**

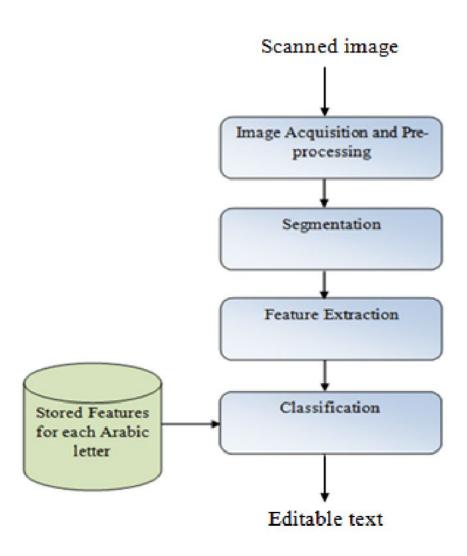
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## i.Project Pipeline

Optical character recognition or optical character reader (OCR) is the recognition process of text obtained from media in the form of typed, handwritten or printed text into machine-encoded text form. The text in question may be presented in the form of a scanned document, a photo of a document, a scene-photo or from subtitle text superimposed on an image.



## ii.Preprocessing Module:

In preprocessing we calculate the skew angle if the scanned image is rotated and fix it and apply Otsu's thresholding method to convert the scanned image to binary image

#### Orignal Image

مسقط العمائية: تشارك واحة المعرفة مسقط اليوم في اجتماعات شبكة كاميردج لمناطق تقنية المعلومات المرامع عقدها بكاميردج بالمملكة المتحدة والتي تستمر لمدة ثلاثة أيام. ويرأس وقد السلطنة الى الاجتماعات سعادة المهتدس على بن مسعود السنيدي وكيل ورارة النجارة والصساعة للتجارة والصساعة للتجارة وأوضح الدارة المؤسسة العامة للمناطق الصناعية الذي غادر البلاد امس والرفد العرافق متوجها الى لتدن. وأوضح المهتدس محمد بن أبو بكر الغساني مدير واحة المعرفة مسقط أحد اعضاء الوفد أن هذه الاجتماعات تهدف الى تبدل الإدارة والمقترحات والتعرف على الخدمات والحوافز التي تقدمها مناطق نقنيات المعلومات وتوفير الى تبدل الإدارة على مجلل استقطاب فرس استثمارية لاعضاء الشبكة .. مبينا أن الواحة تسعى الى اكتساب الخبرات في مجل استقطاب الشركات المعنية من خلال هذه المشاركة. وبين أن الوفد سيعمل على الالتفاء بعدد من المسئولين بمناطق تقنيات المعلومات والمسئولين بالشركات المهتمة بهذا القطاع من أجل النعريف بالحوافز والخدمات التى تقدمها واحة المعرفة مسقط والعمل على استقطاب هذه الشركات والمستثمرين الى السلطنة بشكل عام والواحة بشكل خاص. يذكر أن واحة المعرفة مسقط قد انضمت الى شبكة كاءبردج لمناطق تقنية المعلومات.

#### After Threholding And rotation

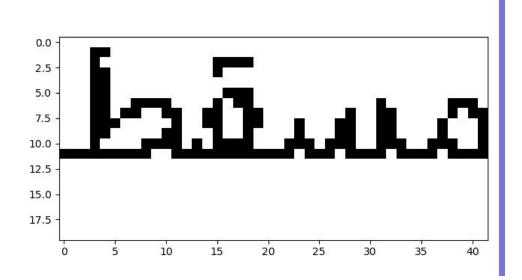
مسعد العدامات تشارك مامة لدعوه مسقط الدور في استاعات سبكه كاسريان ليناطق وسمة الدواوسات المسطنة الله الجوداءات مساحد السيخ عقدة كاسرياج والدعوة الدواوسات الدورج عقدة الاستراخ والتي مستحد التي المستحدة الله المستحدة التي المستحدة المستحدة الله المستحدة المستحدة الله المستحدة المستحددة المستحدة المستحددة المستح

Line Segmentation: we make a horizontal projection and detect the gaps and form the gaps we segment the lines

#### Line(1)

مسقط العمانية: تشارك واحة المعرفة مسقط اليوم في اجتماعات شبكة كامبردج لمناطق تقنية المعلومات

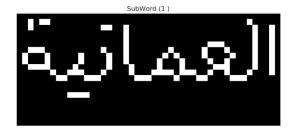
Word Segmentation: after we segment line into separate words by using Vertical projection and deciding on the values between characters

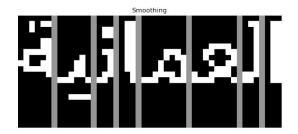


Character Segmentation: First we detect both base line ( making horizontal projections and sum each row and pick the most important one) and we detect the Maximum transition line( we count in each row the transition from black to white and pick the row index with maximum transition)

We use the Maximum transition index and iterate on each col and if the transition is detected

We calculate the start and end index as from black to white and from white to black is my end index and then we choose a cut between them





## iii. Feature Extraction/Selection Module

We choose those features:

- 1. Number of connected components
- 2. Number of holes
- 3. Height to Width ratio
- 4. Max transition columns
- 5. Max transition Row
- 6. White pixels to black pixels ratio
- 7. Segment the image to 4 regions and get the white to the black ratio in each region (ratio 1, ratio 2, ratio 3, ratio 4)
- 8. white pixels in ratio 1/ white pixels in ratio 2
- 9. white pixels in ratio 3 / white pixels in ratio 4
- 10. white pixels in ratio 1 / white pixels in ratio 3
- 11. white pixels in ratio 2 / white pixels in ratio 4
- 12. white pixels in ratio 1 / white pixels in ratio 4
- 13. white pixels in ratio 2 / white pixels in ratio 3

## iv. Model Selection and Training Modules

After we train our model and making a dataset we choose a Gaussian SVM classifier then Neural Networks achieved much better accuracy, As we try to use different SVM classifiers like linear and Non-Linear Polynomial and Gaussian and Amongst the Gaussian kernel and polynomial kernel, we can see that Gaussian kernel achieved a better prediction rate while polynomial kernel misclassified sometimes. Therefore the Gaussian kernel performed slightly better. However, there is no hard and fast rule as to which kernel performs best in every scenario. but due to some runtime issues.

## vi. Performance Analysis Module.

The segmentation performance is almost 80% to 90% through different files

```
Image Word# 571 Text Word#: 571
Appending to training set with accuracy of char segmentation => 82.13660245183888  % Running Time: 6.3249852657318115 for File apr68.png
Image Word# 614 Text Word#: 614
614 614
Appending to training set with accuracy of char segmentation => 80.29315960912052 % Running Time: 5.140742301940918 for File apr69.png
Image Word# 675 Text Word#: 675
Appending to training set with accuracy of char segmentation => 80.5925925925926 % Running Time: 6.288101673126221 for File capr7.png
Image Word# 1867 Text Word#: 1867
1867 1867
Appending to training set with accuracy of char segmentation => 82.11033743974289 % Running Time: 18.20384430885315 for File apr70.png
Image Word# 861 Text Word#: 861
861 861
Appending to training set with accuracy of char segmentation => 84.3205574912892 % Running Time: 8.090716123580933 for File apr71.png
Image Word# 1007 Text Word#: 1007
1007 1007
Appending to training set with accuracy of char segmentation => 81.23138033763655 % Running Time: 10.966155767440796 for File apr72.png
Image Word# 350 Text Word#: 350
```

## **Neural Network**

We achieved an accuracy of 95% on our testset.

```
C:\Users\student\Desktop\Team_14_Cr\OCR-for-Arabic-Scripts\03-Source Code>python edit.py output Expected
capr3.txt: 107
Total distance = 107
Average Accuracy = 95.11%
```

#### **SVM** Gaussian

We achieved an accuracy of 91% on our testset.

```
D:\_Marwan\OCR-for-Arabic-Scripts\OCR-for-Arabic-Scripts\03-Source Code>python edit.py output expected
capr3.txt: 183
Total distance = 183
Average Accuracy = 91.64%
```

## vii. Other developed modules.

#### Classification

We used PyTorch with different architectures to try which one worked best, we used a 5M datapoint dataset, with 2 hidden layers, using the cross entropy loss function. We got an accuracy of 97% on our testset. We trained the project using Microsoft Azure instance, from features extracted from the images. And used Google Colabs to use our training set directly from Drive.

#### viii. Enhancements and Future work.

In our Project, the major obstacle is the segmentation process because of certain characteristics of the Arabian language. This is the stage where the maximum error occurs and therefore as future work we would want to concentrate our work on this stage to attain maximum accuracy in the segmentation phase which deals with the problems of overlapping and characters.

## viv. Team work distribution.

Mohamed Bassel: Character segmentation and integration

Marwan Medhat: Feature Extraction

Mohamed Haitham: Preprocessing, Line Segmentation and Classification(SVM)

Ahmed Khalifa: Word Segmentation and Classification (Neural networks)

## References

- Optical Character Recognition of Arabic Printed Text Safwa Taha, Yusra Babiker, and Mohamed Abbas
- 2. <a href="https://en.wikipedia.org/wiki/Naive\_Bayes\_classifier">https://en.wikipedia.org/wiki/Naive\_Bayes\_classifier</a>
- 3. Printed Arabic Optical Character Recognition using Support vector machine
- 4. <a href="https://stackabuse.com/implementing-svm-and-kernel-svm-with-pythons-scikit-learn/">https://stackabuse.com/implementing-svm-and-kernel-svm-with-pythons-scikit-learn/</a>