The problem	State of the art	Proposed approach	Achieved results

## Combined Radiology and Pathology Classification of Brain Tumors

Rozpoznanie guza mózgu na podstawie obrazu radiologicznego i patologicznego

#### Piotr Giedziun

Supervisor: dr hab. inż. Henryk Maciejewski

4 March 2016



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# Outline

1 The problem

2 State of the art

3 Proposed approach

4 Achieved results



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## Definitions

#### Cancer

Cancer occurs when abnormal cells grow out of control

#### Brain tumor

- Benign or Malignant
- Over time, a low-grade tumor can become a high-grade tumor
- Brain tumors are classified as grade I, grade II, or grade III, or grade IV



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## Brain tumor - Survival rate (5 years or more)

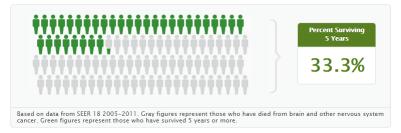


FIGURE - Based on data from SEER 18 2005-2011, cancer.gov



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## Brain tumor - Survival by stage

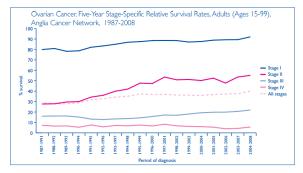
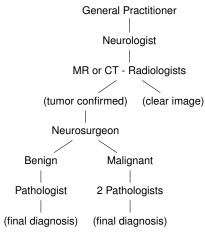


FIGURE – Ovarian cancer, Five-year stage-specific relative survival rates, adults (ages 15-99), Anglia Cancer Network, 1987-2008



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### Brain tumor - Diagnosis process





# **Diagnosis problems**

#### Problems

- Diverse shapes, sizes and appearances of tumors
- Relies on histopathologic examination (biopsy examination)
- Waiting for tests and to start treatment
- Radiology imaging is used only to establish location, size and whether it is benign and malignant tumor

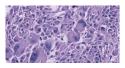


FIGURE - Glioblastoma cells

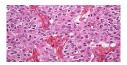


FIGURE - Oligodendroglioma cells



# **Diagnosis problems**

#### Problems

- Diverse shapes, sizes and appearances of tumors
- Relies on histopathologic examination (biopsy examination)
- Waiting for tests and to start treatment
- Radiology imaging is used only to establish location, size and whether it is benign and malignant tumor

#### Targets in the UK

No more than 2 months wait between the date the hospital receives an urgent GP referral for suspected cancer and starting treatment

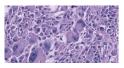


FIGURE - Glioblastoma cells

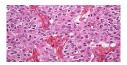


FIGURE - Oligodendroglioma cells



# Aims & Limitations

#### Aims

- Research & build a segmentation mechanism for the MRI scans (ROI selection)
- Research & build a classifier based on the segmented radiological images
- (if possible) Combine the Pathology-based classification with radiology-based classifier

#### Limitations

- Limited access to the MRI samples with the diadnosis provided by the doctor
- Conservative environment only non-black box models



The problem	State of the art	Proposed approach	Achieved results
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### Related work

#### Brain tumor segmentation

- The topic of brain segmentation is relatively popular thanks to BraTS challenge
- Several supervised and unsupervised algorithms were proposed
  - Random Decision Forest that classifies voxels
  - Fuzzy C-means clustering
  - Mean Shift and K-means clustering

#### Brain tumor classification

- Slightly less popular subject (current diagnosis fully rely on histopathology imaging)
- Feature extraction
  - Extraction of structure information
  - Feature selection
- GLCM (Gray-Level Co-occurrence Matrix)



The problem	State of the art	Proposed approach	Achieved results
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### Influential articles





Nitish Zulpe and Vrushsen Pawar GLCM Textural Features for Brain Tumor International Journal of Computer Science, 2012



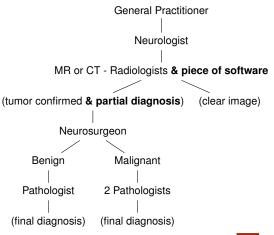
Hassan Khotanlou, Olivier Colliot, and Isabelle Bloch Automatic brain tumor segmentation using symmetry analysis and deformable models

Nationale Superieure des Telecommunications, 2007



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## Brain tumor - Modified diagnosis process



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### Data set

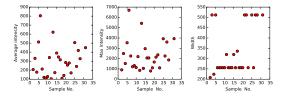


FIGURE - Plots of different attributes of the data set



FIGURE - Viewing angles of MRI scan



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### Data set

#### Summary

- 27 cases with lower grade glioma tumors
- 13 of them with Oligodendroglioma and 14 with Astrocytoma
- Each case has 3 or 4 MRI scans (T1, T1C, FLAIR, and T2)
- Provided samples were taken using different hardware

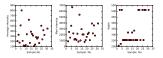


FIGURE - Plots of different attributes of the data set



FIGURE - Viewing angles of MRI scan



The problem	State of the art	Proposed approach	Achieved results
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### Pre-processing

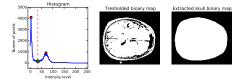


FIGURE - Process of skull extraction

#### FLAIR skull figure



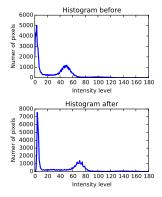


FIGURE - Skulls properties in FLAIR and T2



The problem	State of the art	Proposed approach

# Pre-processing







FLAIR after



FIGURE - Median filter effect on image histogram



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# Segmentation - K-Means

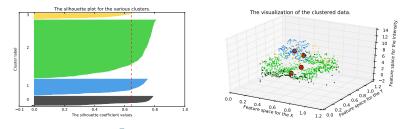


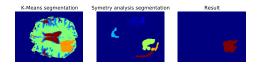
FIGURE - Silhouette analysis for K-Means(k=5)



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# Segmentation - Combined





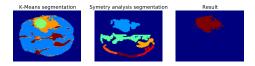
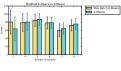


FIGURE - Segmentation with results

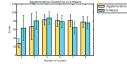


The problem	State of the art	Proposed approach	Achieved results
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# Segmentation - Alternatives

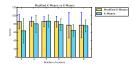






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FIGURE - Agglomerative cluste-







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# Classification

#### Tested methods

- Feature extraction & evaluation
- Texture features extraction with Gray-Level Co-Occurrence Matrix
- Texture features extraction with Local Binary Pattern

#### Classification algorithms

- SVM (Support vector machine)
- Gaussian Naive Bayes
- Logistic Regression
- Random Forest

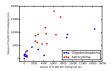


The problem	State of the art	Proposed approach	Achieved results
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## Classification - Feature extraction & evaluation

#### Selected features (out of 59)

- Tumor volume (in mm<sup>3</sup>)
- Tumor position (x,y,z) calculated from the middle of the brain
- Metrics intensity of tumor area
- 8 bins of intensity histogram



#### FIGURE - Selected features extracted from data set

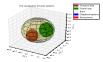


FIGURE – Tumor positional features



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## Classification - Texture features extraction with GLCM & LBP

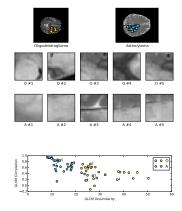


FIGURE - Co-occurence matrix features for Oligodendroglioma and Astrocytoma



The problem	State of the art	Proposed approach	Achieved results ●OO

# Radiology imaging

Tumor segmentation		
METHOD	BEST SCORE	
Mini Batch K-Means (5 clusters) 89.027% (std : 5.408)		
K-Means (5 clusters)	88.168% (std : 5.264)	
K-Means with position (5 clusters)	86.026% (std : 5.282)	
Agglomerative Clustering	88.956% (std : 10.632)	

Cancer classification		
METHOD	BEST SCORE	
Random Forest Classifier 87.000% (std		
Logistic Regression	81.297% (std : 5.744)	
Logistic Regression (texture)	68.285% (std: 0.082)	



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# Combined Radiology and Pathology

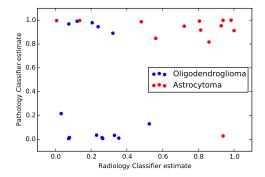


FIGURE – Comparison of Pathology and Radiology results (average estimations of Oligodendroglioma cancer for each sample)



The problem	State of the art	Proposed approach	Achieved results
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### Results

#### Conclusion

- Random Forest classifier validated with k-fold cross validation had average accuracy of 87.0%
- Pre-processing of the input data is a hand-crafted process, that had to be performed
- K-Means had the best score out of Mini Batch K-Means, K-Means with modified input vector (with position), and Agglomerative clustering



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