

UNIVERSITÀ DEGLI STUDI DI NAPOLI FEDERICO II





PATTERN ANALYSIS AND INTELLIGENT COMPUTATION FOR MULTIMEDIA SYSTEMS

Al in healthcare: Activities of the University of Naples Federico II node of the CINI-AIIS Lab

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Artificial Intelligence in Healthcare

Healthcare is considered one of the most promising applications of AI

AI in healthcare can be used to:

- ✓ support patients and physicians
- ✓ transform patient care and administrative processes
- \checkmark analyze patient data to identify potential health risks
- \checkmark help physicians make more accurate diagnoses
- The most common application of traditional ML is precision medicine:
 - ✓ predicting what treatment protocols are likely to succeed on a patient
- DL applications:
 - ✓ Medical images analysis:
 - the aim is to extracting clinically relevant features from images that go beyond what can be perceived by the human eye.
 - ✓ Natural Language Processing:
 - the creation, understanding and classification of clinical documentation
 - Analysis of unstructured clinical notes on patients



AI Applications of the University of Naples Federico II





Multi-task BioNER

- Biomedical Named Entity Recognition (BioNER) involves identifying mentions of biomedical entities from unstructured text
- Developing a BioNER system is still difficult due to the high frequency of synonyms, alternate spellings, polysemous words, and privacy issues.
- We propose TaughtNet a multi-task framework based on knowledge distillation that fine-tunes a single transformer architecture to recognize multiple entity types.



- Methodology:
 - ✓ *Datasets aggregation*. The available single-entity datasets are merged together to build an aggregated multi-entity dataset D_s .
 - ✓ *Retrieval of Teacher predictions*. Each sample in D_s is provided to each teacher as input and the resulting output distributions are stored.
 - ✓ Distributions aggregation. A single output distribution is generated by integrating the output distributions from each teacher
 - ✓ Student Training. A Student is trained by taking both the ground truth and the knowledge of Teachers into consideration.

Kubic FLOTAC Microscope (KFM)

• The Kubic FLOTAC Microscope (KFM) is a compact, low-cost, versatile, and portable digital microscope designed to autonomously analyze faecal specimens prepared with FLOTAC or Mini-FLOTAC.

- The KFM is composed of
 - ✓ electro-mechanical components
 - \checkmark A firmware that allows remote interactions
 - ✓ external agents which enable users to connect with the KFM hardware

The device can be remotely controlled by any user with external devices, like smartphones, tablets or PCs, through a dedicated web interface and app



Al Server for Parasites Egg Detection

• Images collected with the KFM can be stored in the KFM AI server and/or transmitted to diagnostic hubs in order to have a quick diagnosis or a parasitological consultation.



MGMT Promoter Methylation Identification

- Glioblastoma Multiforme (GBM) is known for its extremely low survival rate.
- Alkylating chemotherapy may result inefficient since the O(6)-methylguanine-DNA methyltransferase (MGMT) enzyme repair abilities counter the cytotoxic effects of alkylating agents.
- MGMT promoter regions may be subject to a phenomenon called methylation, a biological process preventing MGMT enzymes from destroying the alkyl agents
- We proposed to select the area of interest in an unsupervised manner, leveraging past medical experience for tumour recognition:
 - ✓ in T1-W slices, tumour areas have pixels whose intensity is higher than cerebrospinal fluids (CSF) but lower than any other kind of tissue
 - ✓ in FLAIR slices, pixels with the highest intensity belong to the tumour region



Convolutional Neural Network for identification

- We chose the FLAIR sequence for the good performance shown in the literature
- We adopted a Convolutional Neural Network (CNN) to face the task of MGMT promoter methylation identification.
 - We built a sequential network from the ground called MGMTClassifier composed of seven Convolutional Blocks and two fully connected layers separated by Rectified Linear Unit (ReLU) as an activation function.
- To reduce the number of training parameters and avoid overfitting at the same time, we adopt depth-wise separable convolution
- This simple but effective DL-based approach is able to outperform state-of-the-art solutions while consisting of less than 0.29% of their parameters (about 10 million of typical CNNs versus 40561 of the proposed approach).



Assessing Brain Health with the Brain-Age Paradigm

- Machine Learning methods are used to model chronological age as a function of structural brain MRI scans in healthy people
- The extent to which each subject deviates from healthy brainageing trajectories, expressed as the difference between predicted and chronological age (brain-predicted age difference, brain-PAD), has been proposed as an index of structural brain health, sensitive to brain pathology in a wide spectrum of neurological and psychiatric disorders.
- We applied the brain age paradigm to a target clinical population of patients with Fabry Disease (FD), a rare genetic multisystemic disorder that also involves the brain but lacks quantitative neuroimaging biomarkers.
- Our brain-age model was based on the DenseNet architecture



Axillary Lymph Node Status Assessment in Breast Cancer

- Breast cancer (BC) is the most diagnosed cancer among women worldwide
- Axillary lymph nodes metastatic involvement, is one of the most important prognostic factors
- Axillary lymph nodes histological exam is the gold standard to determine their involvement in the BC patients:
 - ✓ It is an invasive procedures with complications
- The research activity is in collaboration with Università Campus Bio-Medico of Rome
- Multimodal dataset:
 - ✓ Different Magnetic Resonance Imaging (MRI) sequences:
 - Dynamic Contrast Enhanced (DCE)
 - Diffusion Weighted Imaging (DWI)
 - ➢ T2 Weighted
 - $\checkmark~$ Clinical and Histological features



Multimodal Approach



Deep Feature representation for CBIR in smart PACS

- Picture Archiving and Communication Systems (PACSs) represent an actual possibility to archive and organize the growing amount of data in Pathological Anatomy
- The Content-Based Image Retrieval (CBIR) methodology can be involved in the PACSs using a query-by-example task.
 - ✓ One of many crucial points of CBIR concerns the representation of images as feature vectors
 - Our study explored different representations by features extracted from pretrained CNNs:
 - ✓ We valuated features extracted from different layers using different dimensionality reduction techniques.
 - ✓ we analyzed intend to understand which dataset category is not correctly retrieved.



Thank you for your attention!