Motivation



Critical diseases in China (25 types) mia-reperfusion injury Glomeruli Collectin robial contamina Damageo stilue network glomerulu Damaged tubule Donor-derive Normal Diseased dendritic cell kidnev kidney maturation OUNDATION FOR MEDICAL EDUCATION AND RESEARCH, ALL Damag Decreased blood flow Normalized blood flow **Heart Attack**

Malignant tumor, acute myocardial infarction, sequelae of cerebral apoplexy, major organ transplantation or hematopoietic stem cell transplantation, coronary artery bypass grafting (or coronary artery bypass grafting), end-stage renal disease (or uremia phase of chronic renal failure)



COVID-19-Chest-X-ray-Detection: classify COVID-19, Viral pneumonia and Normal Chest X-ray images with an accuracy of 98.3%

50



<u>Covid19model</u>: modelling estimated deaths and cases for COVID19





xDNN-SARS-CoV-2-CT-Scan: an eXplainable Deep Learning approach (xDNN)



Figure 1. This figure illustrates the layered architecture of the proposed method. It has the form of a deep neural network but is using clear to understand prototypes (actual images). The density layer identifies the local peaks of the density and empirically derived probability distributions. The prototypes are actual training data samples (in this case, images) which are highly representative (local peaks of the density and empirically derived probability distributions.



COVID-19-CT-Seg-Benchmark: Accurate segmentation of lung and infection in COVID-19 CT





<u>COVID-Net</u>: a convolutional neural network for detecting COVID-19 through chest radiography





Health Encyclopedia Code (HEC)



Infectious diseases



Type III: influenza, mumps..

Ranked: Global Pandemic Preparedness by Country





HEC in other countries



| Country | PHR system | Description | Data content |
|-----------|--------------------|--|---|
| US | Blue Button | patients view online and download their own personal health records. | |
| UK | PatientView App | Renal patients, Renal patient view, provides online information for kidney patients' including diagnoses, treatment and latest test results. | Comment C |
| Australia | PCEHR | Personally Controlled Electronic Health Records | And |



covid-cxr: Neural network model for classifying chest X-rays by presence of COVID-19 features



Fig 1. Flow diagram. We collected a dataset of 3506 patients with chest CT exams. After exclusion, 3,322 eligible patients were included for the model development and evaluation in this study. CT exams were extracted from DICOM files. The dataset was split into a training set (to training the model), and the independent testing set at the patient level. A supervised deep learning framework (COVNet) was developed to detect COVID-19 and community acquired pneumonia. The predictive performance of the model was evaluated by using an independent testing set. COVNet = COVID-19 detection neural network.

HEC in other countries



Government-led

National initiative:

e.g., Australia, Italia



NGO-led

"Service-oriented":

e.g., US, UK

U.S. HEALTH CARE RANKS LAST AMONG WEALTHY COUNTRIES

A recent international study compared 11 nations on health care quality, access, efficiency, and equity, as well as indicators of healthy lives such as infant mortality.



Spurce, K. Davis, K. Stremikis, D. Squives, and C. Schoam, Mimor, Mimor on the Wall How the Performance of the U.S. Health Care System Compares Internationally, 2014 Update, the Commonwealth Fund, June 2014.



Enforcement

State financial support

e.g., Denmark



HEC in other countries



| Country | PHR system | Description | Data content |
|---------|--------------------------------|--|---|
| Denmark | sundhed.dk | Danish Healthcare Services enables citizens and healthcare professionals | Bur sundheid, dk Din sundheid, dk Din sundheidspontal Din sun |
| Italia | Lombardy CRS-SISS system | Lifelong PHR embedded in the regional healthcare information system of Lombardy | Image: Control of the second and th |

Content of HEC



String 1 : Personal Information

String 2 : Omics

- **String 3 : Physical Examination Information**
- **String 4 : Wearable Device**
- **String 5: Medical Treatment Information**

- Infectious diseases:
 - string 1+2+3+5;
- Basic chronic diseases:
 - string 1+2+4;
- Critical disease:
 - string 1+2+5



String 1 : Personal information







Personal information: name, gender, nationality, ID, personal biometric information, etc. Personal track: Public security, mobile operation network, health commission, traffic (high-speed railway, airport, expressway crossings)







Omics: analysis of complete genetic or molecular profiles. Genetics: single genes Genomics: all genes (genomes) and their relationship.

String 3 : physical examination



THE DRAMMATION CONTRACTORY CONTRACTORY



height, weight, blood pressure, blood sugar, blood routine, urine routine, diet, sleep

String 4 : Wearable device







Sports and fitness outdoor areas: heart rate, cadence, air pressure, diving depth, altitude, sleep, monitoring running, jogging, walking, cycling, skiing, weightlifting repeatability count

Healthcare: blood pressure, heart rate, blood oxygen, pulse

String 5 : medical treatment information



Text:

History of basic medical, history of family, history of allergies, history of immunizations, medical records, medication use, surgical and radiotherapy procedures, laboratory results, diagnostic reports



Images and videos:

X-ray, CT, Ultrasound, MRI, PET, pathology



Content of HEC



Data storage:

- Hospitals: basic data (text + image)
- Cloud brain of PCL: features (NLP + image + video)
- **Computation:** Cloud brain of PCL





Security



□ Against attack :

Photo copy

GAN: Generate the same HEC of others

Encryption:

Fragile watermark









- D Physical attack :
 - system IntrusionData tampering





| | Project name | content | Organization |
|---------------------------|-------------------------|--|---------------------------------------|
| COVID-19 cases and deaths | COVID-19 | COVID-19 cases and deaths in the world | Johns Hopkins University |
| | covid-19-data | COVID-19 cases and deaths in the US | The New York Times |
| | covid19india-react | COVID-19 cases and deaths in India | Individual |
| | COVID-19 | COVID-19 cases and deaths in Italia | Individual |
| | covid-19-data | COVID-19 confirmed cases, deaths, and tests in the world | Our World in Data |
| Radiological images | covid-chestxray-dataset | COVID-19 cases with chest X-ray or CT images | University of Montreal |
| | COVID-CT | 349 CT images containing clinical findings of COVID-19 from 216 patients | University of California at San Diego |
| | covid19 | COVID-19 CT segmentation dataset | Unknow |



| | Project name | content | Organization |
|---------------------|----------------------------|--|---|
| Radiological images | <u>COVID-CT</u> | 349 CT images containing clinical findings of COVID-19 from 216 patients | University of California at San Diego |
| | <u>covid19</u> | COVID-19 CT segmentation dataset | Unknow |
| | COVID-19-CT-Seg-Benchmark | COVID-19 CT scans from non-COVID-19 CT scans | Unknow |
| | covid-19-image-repository | Radiological images about COVID-19 | Unknow |
| | COVID-19 chest xray | A database of COVID-19 cases with chest X-ray or CT images. | Kaggle |
| | SARS-COV-2 Ct-Scan Dataset | A SARS-CoV-2 CT scan dataset, containing 1252 CT scans that are positive for SARS-CoV-2 infection (COVID-19) and 1230 CT scans for patients non-infected by SARS-CoV-2 | Kaggle |
| Natural language | <u>covid-qa</u> | A collection of COVID-19 question-answer pairs and transformer baselines for evaluating QA models | Unknow |



covid19_scenarios: Models of COVID-19 outbreak trajectories and hospital demand

| POPULATION (?) | | | EPIDEMIOLOGY 🕜 | | |
|---------------------------|--------------------|-----|----------------------------|-------------|----|
| Population 2 | 327167434 | \$ | Annual average Ro 2 | 4.08 2 4.98 | • |
| Age distribution | United States of | 400 | Latency [days] | 3 | |
| Initial number of cases 🧃 | 1 | \$ | Infectious period [days] 👔 | 3 | : |
| Imports per day 👔 | 0.1 | \$ | Seasonal forcing (? | 0 | • |
| Hospital Beds (est.) 🧃 | 798288 | \$ | Seasonal peak ? | January | × |
| ICU/ICMU (est.) ? | 49499 | • | Hospital stay [days] 👔 | 3 | : |
| Confirmed cases (?) | United States of . |) × | ICU stay [days] | 14 | • |
| Simulation time range ? | 08 Feb 2020 | | Severity of ICU overflow ? | 2 | .: |
| annamentange () | 31 Aug 2020 | | | | - |
| Number of runs 🤤 | 15 | | | | |