## Data science: Case study

#### Problem statement

- □**Big data**: The volume of this dataset is around 43GB. This is make it big data problem.
  - Chest X-Ray dataset has 112120 gray scale images with 1024 by 1024 image size.
- ☐ Multiple class with multiple labels:
  - The dataset contains x-ray images that shows one or more Thorax Disease and the total number of diseases is 14.
  - This make the problem as multiple class and multiple labels problem.

## Challenges

- Big data: with 42 GB of images
- Data types: images and text thus NLP is required to extract labels
- Multiple class and multiple labels
- ☐ Requirements powerful hardware : Hadoop
- Project management tools
  - Confluence
  - JIRA
  - BITBUCKET

#### How Data are store

#### ☐ Data Description:

- Stored in CSV file which contains information about each patients and what diseases he/she has. This shown here
- xray data statistics+.html

#### ☐ Most important Columns :

- Image index : unique name of each image
- Finding Labels: List of all possible disease in each X-ray image

#### ■Where is the labels:

- NLP is used to extract the labels from the finding labels.
- Convert images final jan29.html
- ☐ Images are stored in sperate folder with unique names

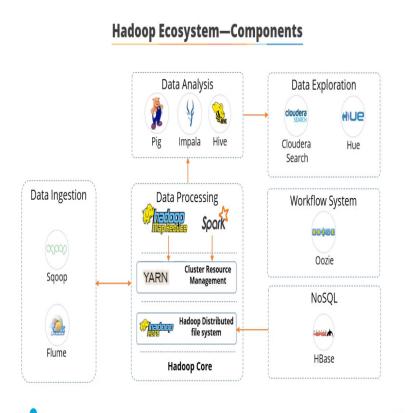
#### Dataset: Classes distribution

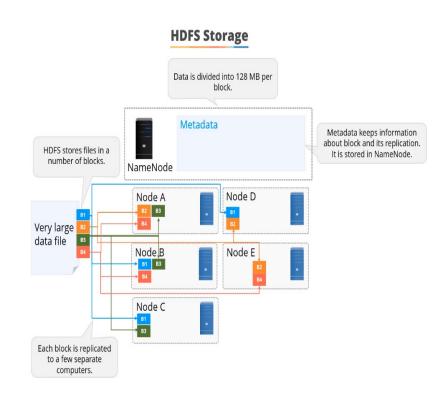
TABLE I
CHEST X-RAY CLASS DISTRIBUTION

| Class Name         | Frequency | Percentage |
|--------------------|-----------|------------|
| Atelectasis        | 11559.0   | 10.309490  |
| Cardiomegaly       | 2776.0    | 2.475919   |
| Effusion           | 13317.0   | 11.877453  |
| Infiltration       | 19894.0   | 17.743489  |
| Mass               | 5782.0    | 5.156975   |
| Nodule             | 6331.0    | 5.646629   |
| Pneumonia          | 1431.0    | 1.276311   |
| Pneumothorax       | 5302.0    | 4.728862   |
| Consolidation      | 4667.0    | 4.162504   |
| Edema              | 2303.0    | 2.054049   |
| Emphysema          | 2516.0    | 2.244024   |
| Fibrosis           | 1686.0    | 1.503746   |
| Pleural Thickening | 3385.0    | 3.019087   |
| Hernia             | 227.0     | 0.202462   |
| No Finding         | 60361.0   | 53.836068  |

Imbalanced dataset

## Requirements: Hardware infrastructure





## Requirements: Why Hadoop

#### **Challenges of Traditional System**

In traditional system, storing and retrieving volumes of data had three major issues, such as:

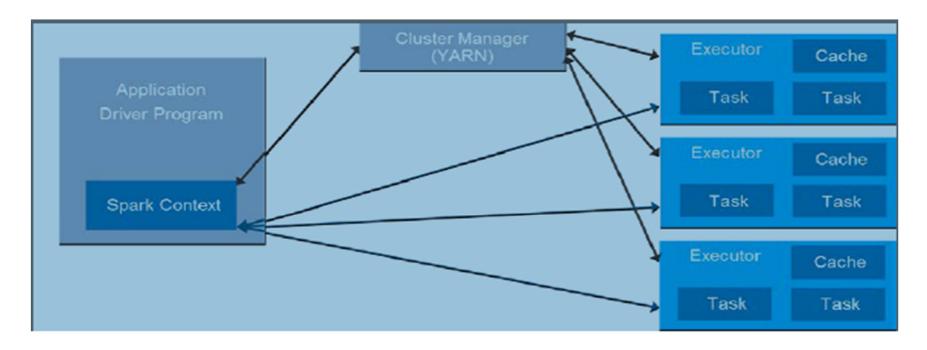
# Cost Speed Reliability \$10,000 to \$14,000, per terabyte and analysis Speed Reliability Difficulty to fetch data

#### **Need for HDFS**

HDFS resolves all the three major issues of the traditional file system.



## Parallel processing: Task distribution



Number of executors: 16 Number of cores: 46

Batch size: 1024, 512, 2048..etc

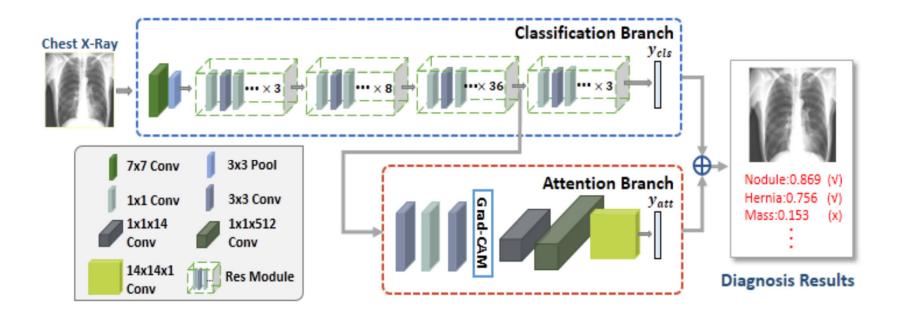
## Deep learning using transfer learning

- ☐ Transfer learning is utilized to predict the diseases in images using
  - Resent
  - Densenet
  - VGG16
  - Inception

☐ The complete code is here:

HTML code Chest ray.html

## Deep learning: workflow



Number of layers: 227 layers

Last layer: sigmoid

### Our Results

TABLE II
OUR RESULTS IN COMPARSION WITH OTHER STATE OF THE ART RESULTS.

| Class Name         | OurAccuracy    | Wang et al(2017) | Yao et al. (20 |
|--------------------|----------------|------------------|----------------|
| Atelectasis        | 0.794554080842 | 0.716            | 0.772          |
| Cardiomegaly       | 0.873621830242 | 0.807            | 0.904          |
| Effusion           | 0.867429446012 | 0.784            | 0.859          |
| Infiltration       | 0.69914106984  | 0.609            | 0.695          |
| Mass               | 0.815454190407 | 0.706            | 0.792          |
| Nodule             | 0.720415340566 | 0.671            | 0.717          |
| Pneumonia          | 0.722111574916 | 0.633            | 0.713          |
| Pneumothorax       | 0.87348621061  | 0.806            | 0.841          |
| Consolidation      | 0.784385714286 | 0.708            | 0.788          |
| Edema              | 0.891658152995 | 0.835            | 0.882          |
| Emphysema          | 0.884076076811 | 0.815            | 0.829          |
| Fibrosis           | 0.78989952945  | 0.769            | 0.767          |
| Pleural Thickening | 0.759304768164 | 0.708            | 0.765          |
| Hernia             | 0.740815766503 | 0.767            | 0.914          |
| Average auc        | 0.80117        | 0.73814          | 0.80271        |