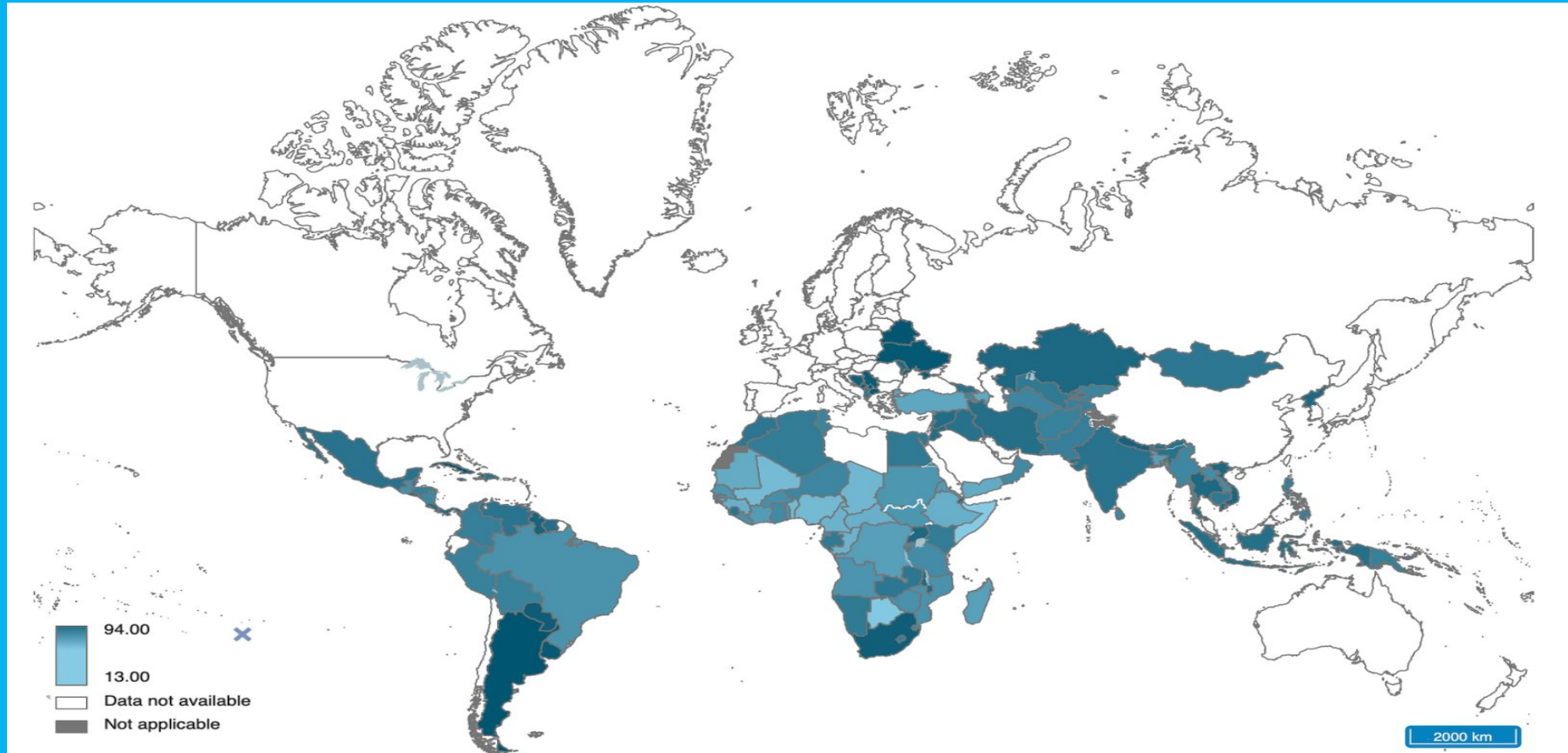




Diagnosing Pneumonia with AI

Children aged under 5 years with pneumonia symptoms taken to a healthcare provider (%)



Motivation

AI is a **powerful tool**. If we can **improve healthcare** services, we should.

Image recognition is just one area of AI but is being used in analysis of x-rays and CAT scans with **great success**.^[1,2]



[1] McKinney, S.M., Sieniek, M., Godbole, V. et al. International evaluation of an AI system for breast cancer screening. *Nature* 577, 89–94 (2020) doi:10.1038/s41586-019-1799-6

[2] B. Parmadean et al. Transfer Learning from Chest X-Ray Pre-trained Convolutional Neural Network for Learning Mammogram Data, (Procedia Computer Science **Volume 135**, 2018, Pages 400-407)

<https://doi.org/10.1016/j.procs.2018.08.190>

808,694

Number of deaths worldwide in
children aged under 5 years (2017)

What is Pneumonia?

Lungs are made up of small air filled sacs, called alveoli.

Pneumonia causes these alveoli in the lungs to fill up with fluid.

This causes them to become enflamed.





“

**Pneumonia is the #1
infectious-related cause of
death in developed countries**

”

Diagnosis & Treatment

Main types of Pneumonia:
Bacterial, Viral.

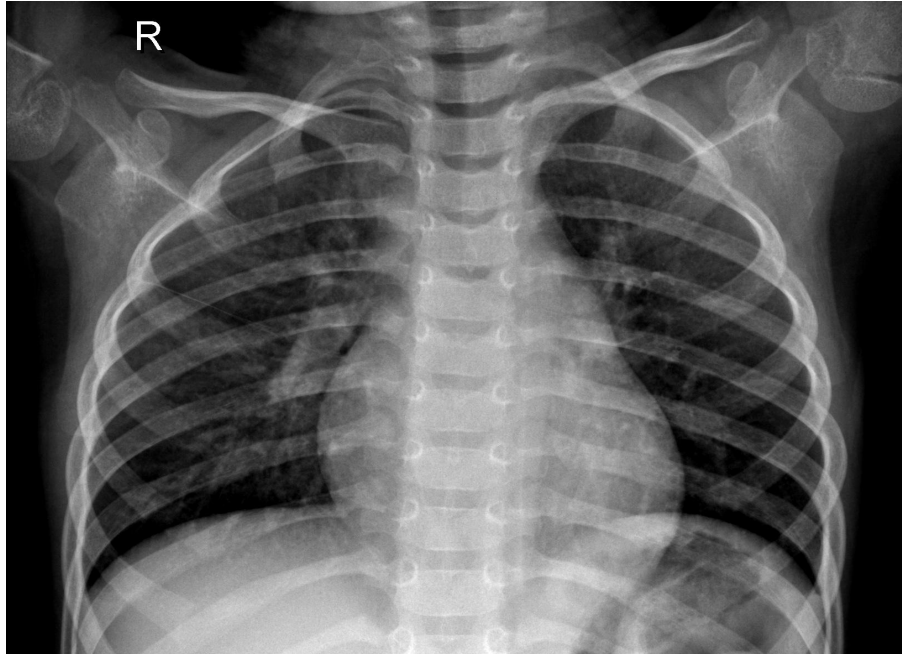
Differential diagnoses only discovered
through failure of treatment.

↑ Diagnoses Time = ↑ Mortality Rate



AI for Image Recognition

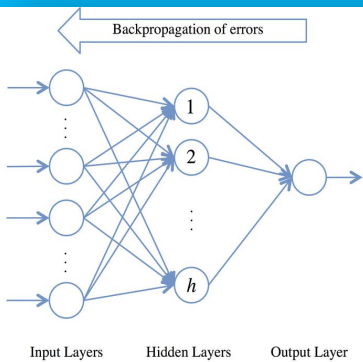
Can you tell the difference?



Healthy



Pneumonia



Neural Networks Can

98% Sick Patients

Correctly Diagnosed

86% Healthy Patients

Correctly Diagnosed

93%

Overall Accuracy

Benefits

- Saves time and resources.
- Higher overall accuracy than traditional diagnosis methods.
- Potential for diagnosing other diseases.

We can improve diagnoses speed and reduce mortality rates.

Thanks!
Any questions?

Credits

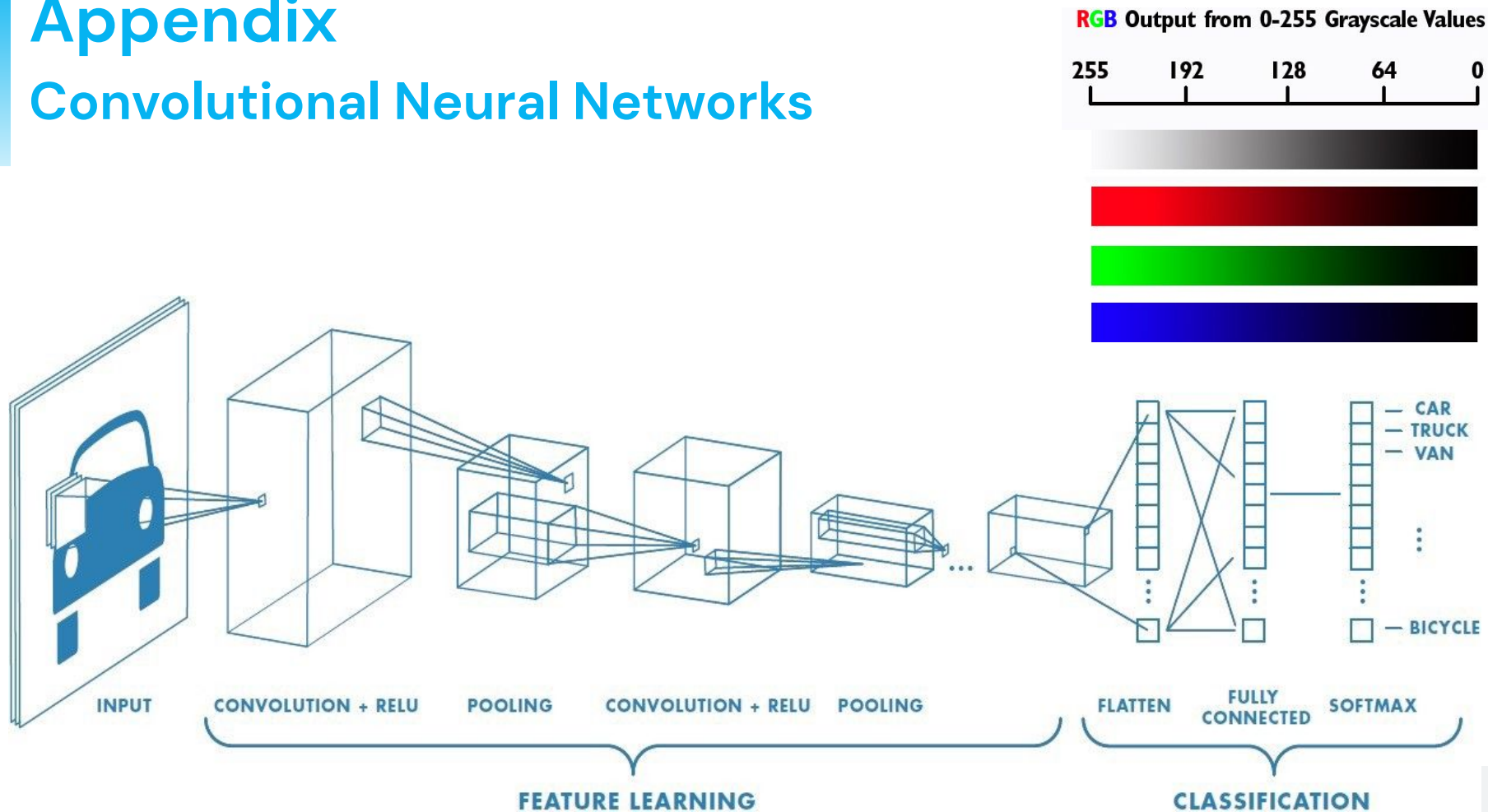
Special thanks to all the people who made and released these awesome resources for free:

- Presentation template by [SlidesCarnival](#)
- Photographs by [Unsplash](#)

Find the notebook at: https://github.com/Greg-S-12/pneumonia_cnn

Appendix

Convolutional Neural Networks



Appendix

The Model:

```
model_2 = Sequential()

model_2.add(VGG16(include_top=False, input_shape=(224,224,3)).layers[0])
model_2.add(VGG16(include_top=False, input_shape=(224,224,3)).layers[1])
model_2.add(VGG16(include_top=False, input_shape=(224,224,3)).layers[2])
model_2.add(VGG16(include_top=False, input_shape=(224,224,3)).layers[3])

model_2.add(Conv2D(filters=64, kernel_size=(3,3), padding="same", activation="swish", name='Conv1_1'))
model_2.add(Conv2D(filters=64, kernel_size=(3,3), padding="same", activation="swish", name='Conv1_2'))
model_2.add(MaxPool2D(pool_size=(2,2),strides=(2,2), name='Pool1'))

model_2.add(Conv2D(filters=128, kernel_size=(3,3), padding="same", activation="swish", name='Conv2_1'))
model_2.add(Conv2D(filters=128, kernel_size=(3,3), padding="same", activation="swish", name='Conv2_2'))
model_2.add(MaxPool2D(pool_size=(2,2),strides=(2,2), name='Pool2'))

model_2.add(Conv2D(filters=128, kernel_size=(3,3), padding="same", activation="swish", name='Conv3_1'))
model_2.add(Conv2D(filters=128, kernel_size=(3,3), padding="same", activation="swish", name='Conv3_2'))
model_2.add(MaxPool2D(pool_size=(2,2),strides=(2,2), name='Pool3'))

model_2.add(Conv2D(filters=256, kernel_size=(3,3), padding="same", activation="swish", name='Conv4_1'))
model_2.add(BatchNormalization())
model_2.add(Conv2D(filters=256, kernel_size=(3,3), padding="same", activation="swish", name='Conv4_2'))
model_2.add(BatchNormalization())
model_2.add(MaxPool2D(pool_size=(2,2),strides=(2,2), name='Pool4'))

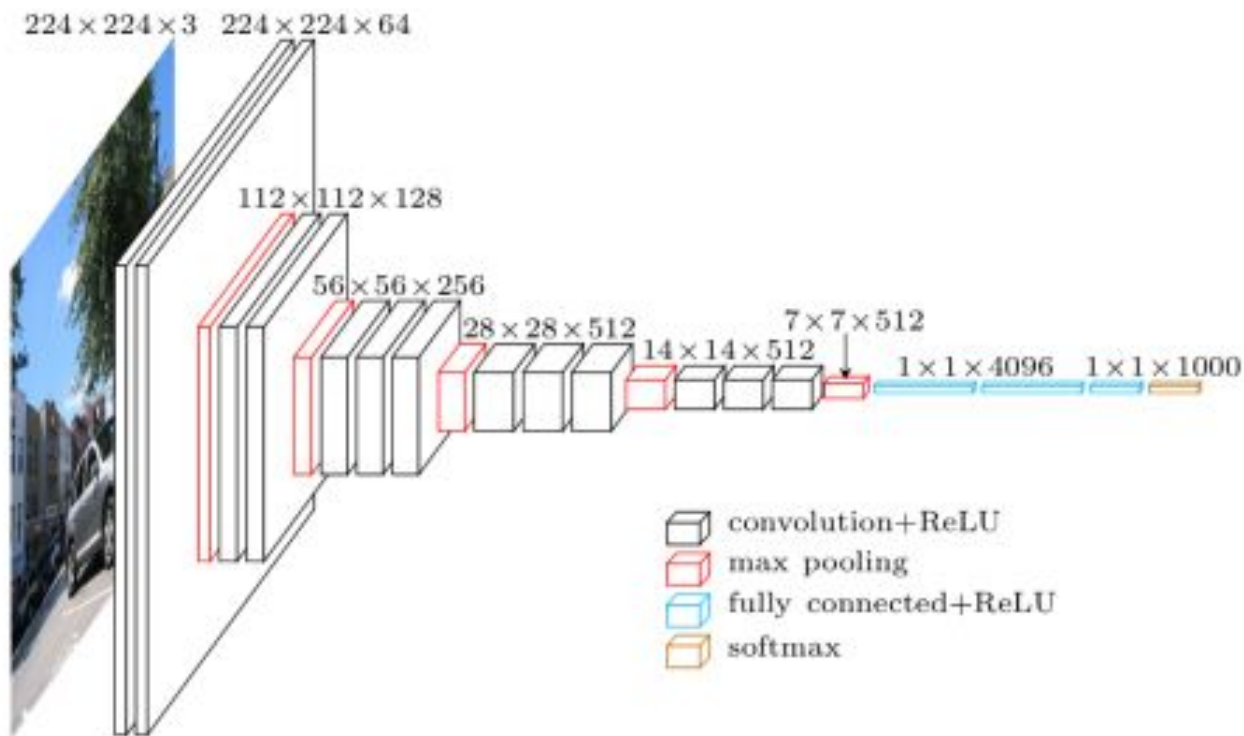
model_2.add(Conv2D(filters=256, kernel_size=(3,3), padding="same", activation="swish", name='Conv5_1'))
model_2.add(BatchNormalization())
model_2.add(Conv2D(filters=256, kernel_size=(3,3), padding="same", activation="swish", name='Conv5_2'))
model_2.add(BatchNormalization())
model_2.add(MaxPool2D(pool_size=(2,2),strides=(2,2), name='Pool5'))

model_2.add(Conv2D(filters=512, kernel_size=(3,3), padding="same", activation="swish", name='Conv6_1'))
model_2.add(BatchNormalization())
model_2.add(Conv2D(filters=512, kernel_size=(3,3), padding="same", activation="swish", name='Conv6_2'))
model_2.add(BatchNormalization())
model_2.add(MaxPool2D(pool_size=(2,2),strides=(2,2), name='Pool6'))

model_2.add(Flatten(name="Flatten"))
model_2.add(Dense(units=1024,activation="swish", name='Dense1'))
model_2.add(Dense(units=512,activation="swish", name='Dense2'))
model_2.add(Dense(units=2, activation="softmax", name='Result'))
```

Appendix

Transfer Learning: VGG16



Appendix

Gradient computation: Backpropagation algorithm

Intuition: $\delta_j^{(l)}$ = "error" of node j in layer l .

For each output unit (layer $L = 4$)

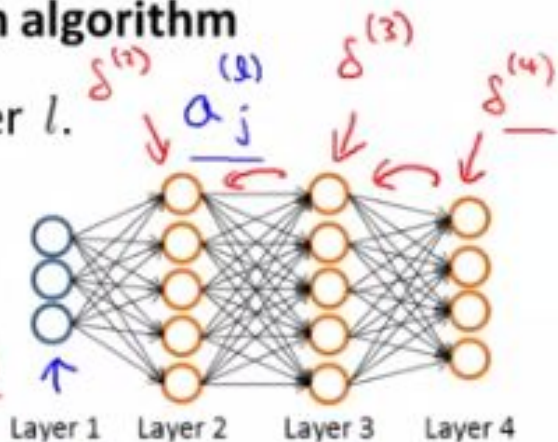
$$\delta_j^{(4)} = a_j^{(4)} - y_j$$

$$(h_{\theta^{(4)}})_j \quad \delta_j^{(4)} = a_j^{(4)} - y_j$$

$$\delta^{(3)} = (\Theta^{(3)})^T \delta^{(4)} \cdot g'(z^{(3)})$$

$$\delta^{(2)} = (\Theta^{(2)})^T \delta^{(3)} \cdot g'(z^{(2)})$$

$$(No \quad \delta^{(1)}) \quad 2$$



Courtesy of Andrew Ng (Coursera)