

# Dermatology



Algorithm

Cancerous or not

whether a region of skin  
tissue is cancerous or not.

Images



...



Labels

Cancerous

Non-cancerous

Convolutional  
Neural  
Network

129,000 images

such an algorithm in the course.



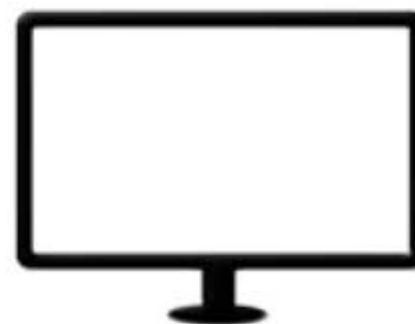
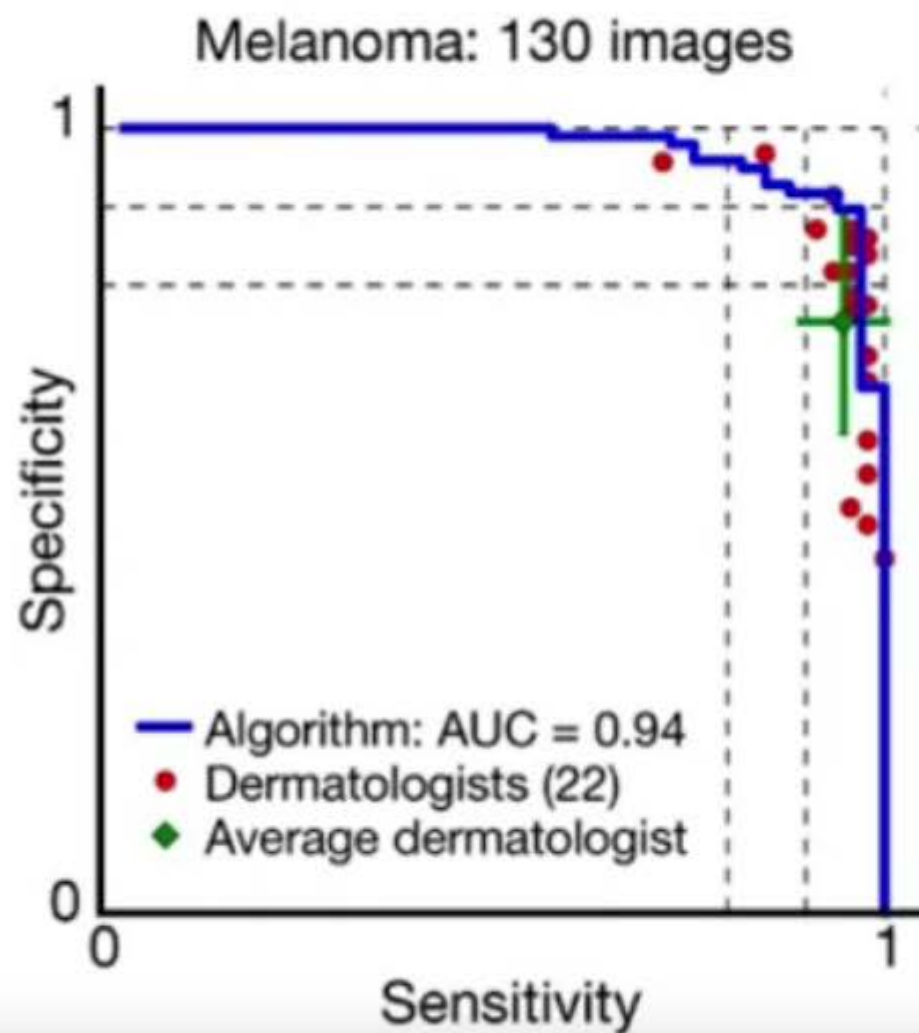
→ Cancerous or not



versus



on a new set of images.



versus



In this study, it was found that the algorithm



1:43 / 2:08



# Ophthalmology



→ Diabetic Retinopathy (DR)



**Retinal Fundus Photos**

Our second example  
is an ophthalmology,



Images



Labels

DR

30%

...



No DR

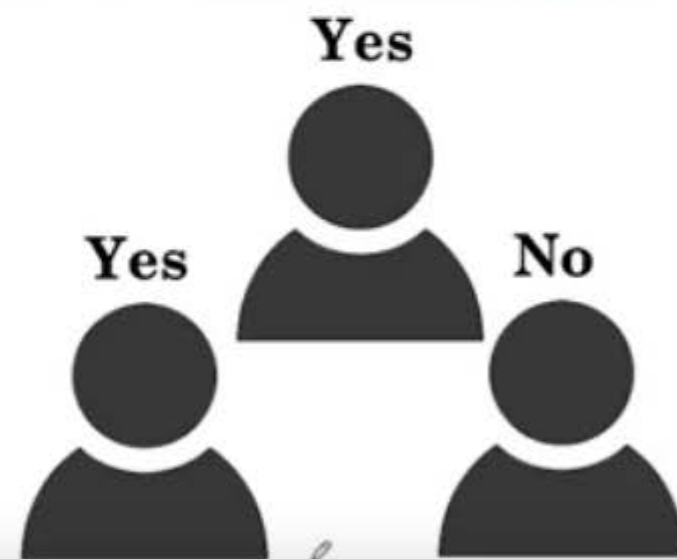
Convolutional  
Neural  
Network

128,000 images

We'll see some methods for  
tackling this challenge.

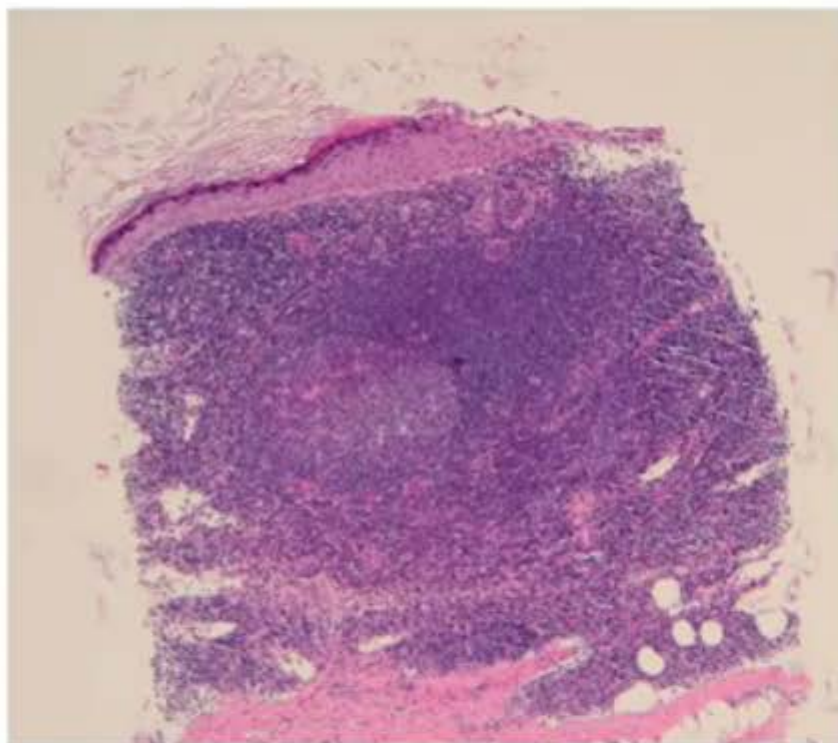


Comparable to



In the study, a majority vote of

# Histopathology



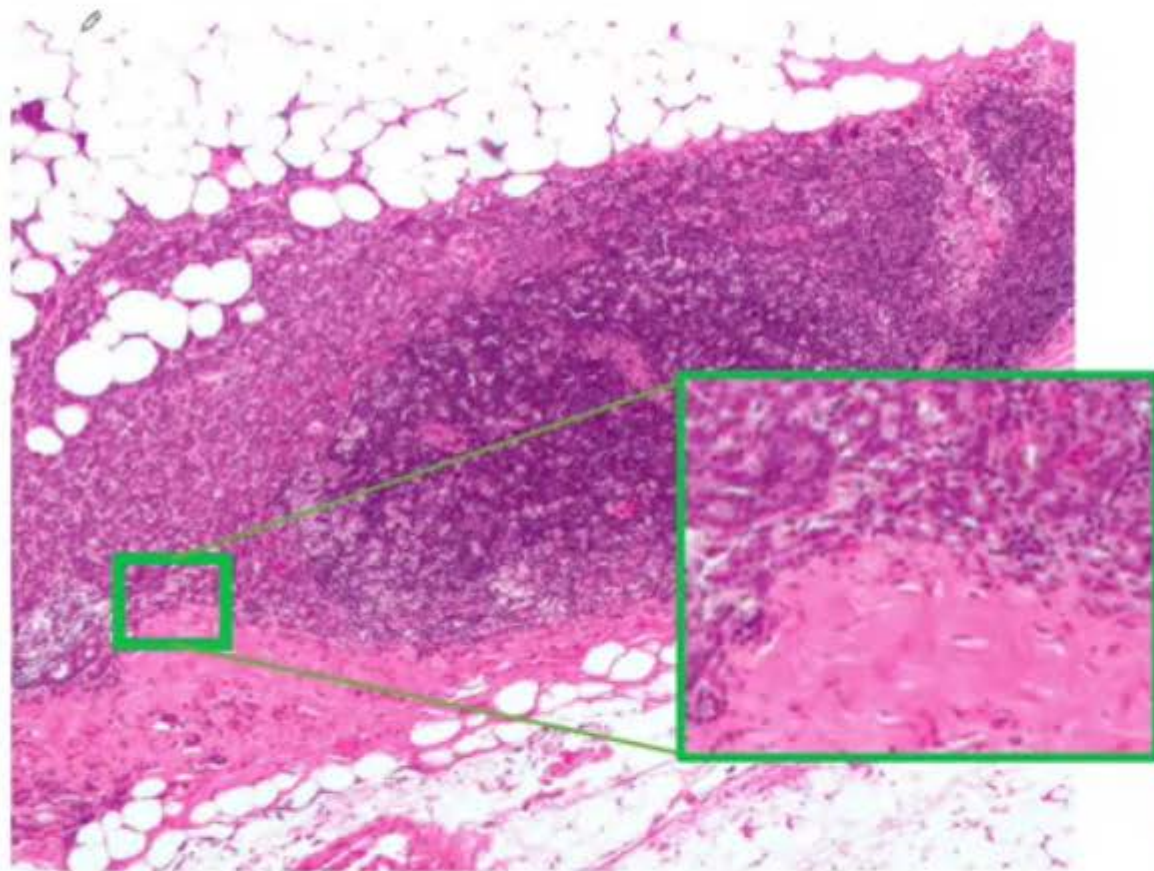
Whole-slide image

→ Cancer Spread  
To Lymph Nodes?

and the chance of recovery.

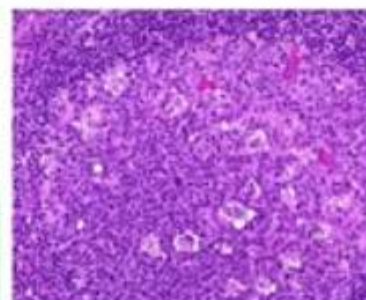


# Histopathology

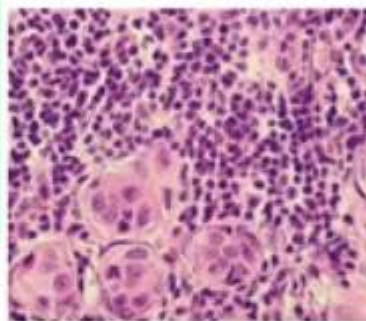


Images

Labels



Not Spread



Spread

270 whole-slide images

500,000+ patches

that instead of  
feeding in one large,

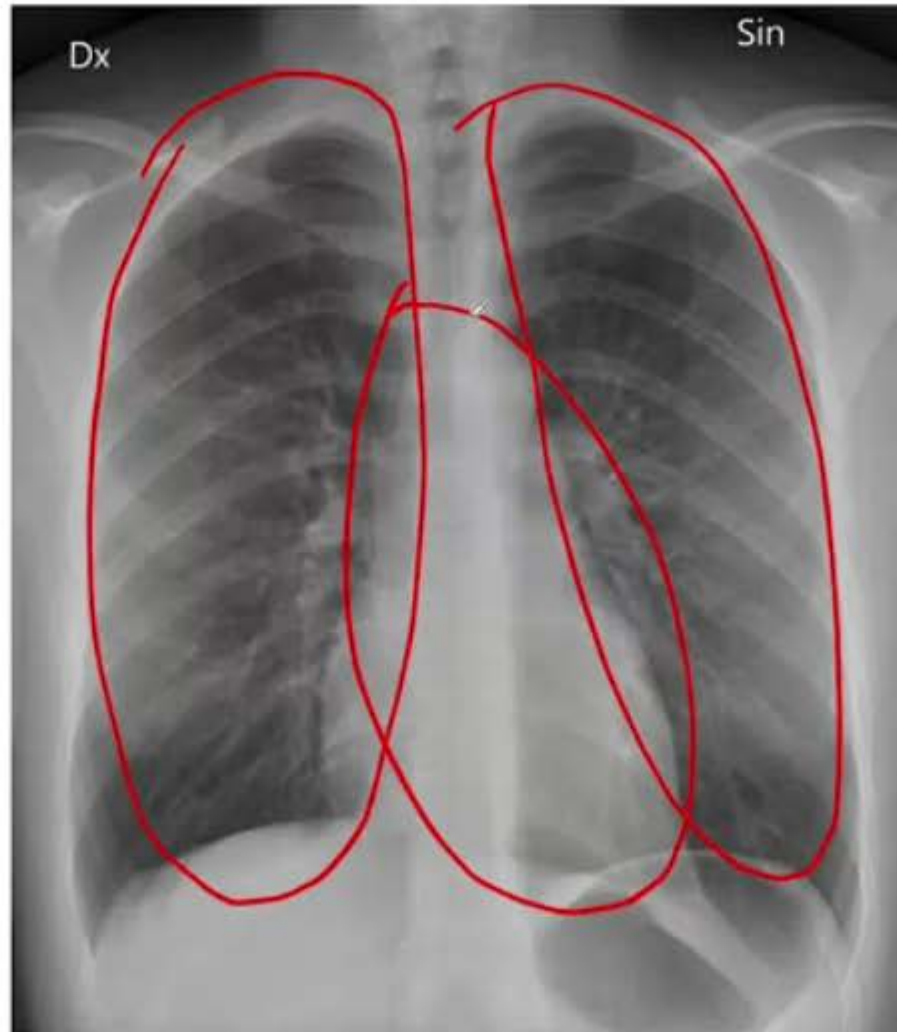
1.00

**2 billion  
per year**



in medicine with about 2 billion chest  
X-rays that are taken for a year.

Critical for detection  
Of pneumonia, lung  
cancer etc



if a patient has pneumonia or  
lung cancer or another condition.

1.00

Mass



Normal



Play

three chest X rays that are normal.



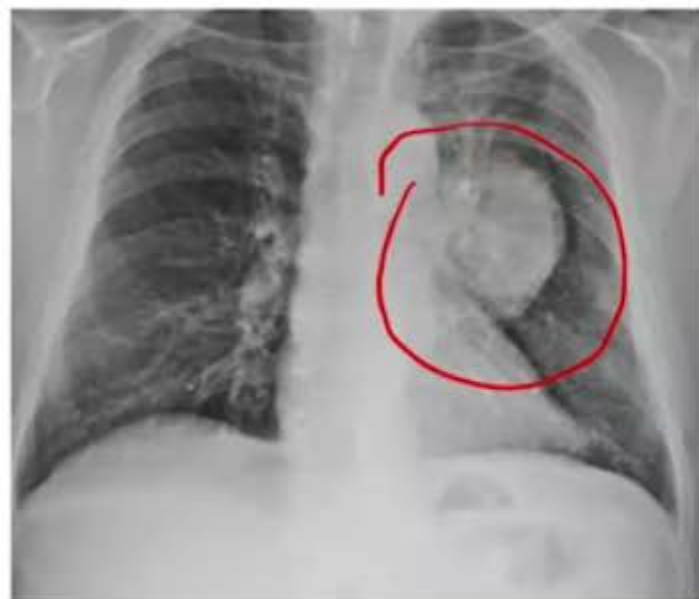
1:36 / 2:30

deeplearning.ai





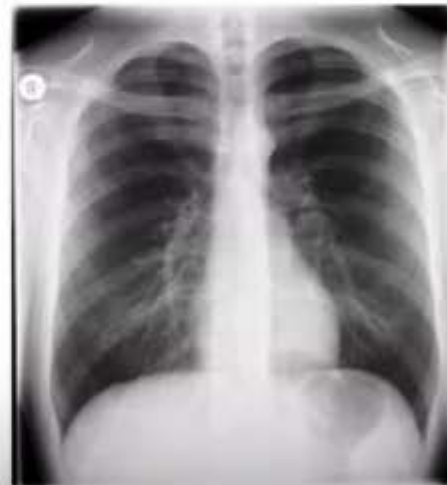
1.00



Mass?



Share



Play

For our own reference, a mass is defined as a lesion or in other words

2:16 / 2:30





1.00

Train

Images



Algorithm

Learn

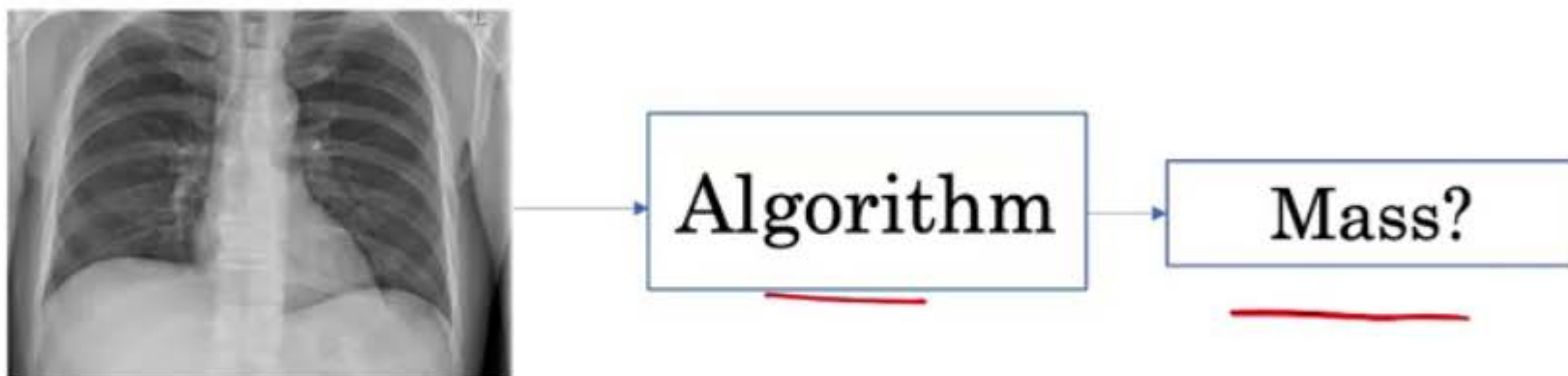
Labels

Mass

Normal

1.00

[Share](#)



deep learning algorithm  
model  
neural network  
convolutional neural network

Play



0:32 / 1:52

deeplearning.ai



1.00

Train

Images



Algorithm

Update

Prediction Probabilities

Share

Desired Label: 1 (Mass)

0.48

Error (Loss) = 0.32

Error (Loss) = 0.31

0.51

Desired Label: 0 (Normal)



1:30 / 1:52



1.00

Train

Images



Algorithm

Updated

Prediction Probabilities

Desired Label: 1 (Mass)

0.60

Error (Loss) = 0.22

Error (Loss) = 0.15

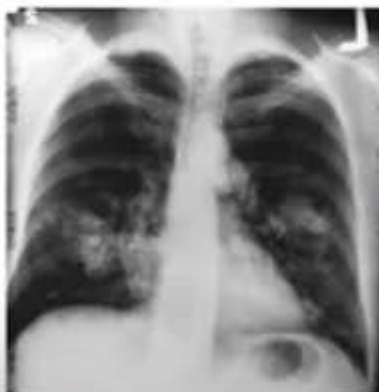
0.30

Desired Label: 0 (Normal)



Algorithm

Penguin?



Algorithm

Mass?

Mass?



0:43 / 1:38





### 3 Key Challenges

Class Imbalance

Multi-Task

Dataset Size

Weighted Loss  
/ Resampling

Multi-Label  
Loss

Transfer  
Learning + Data  
Augmentation

## Class Imbalance



**Normal**



**Normal**



**Mass**



**Normal**

prevalence

1.00



Algorithm

Poor Learning

Normal

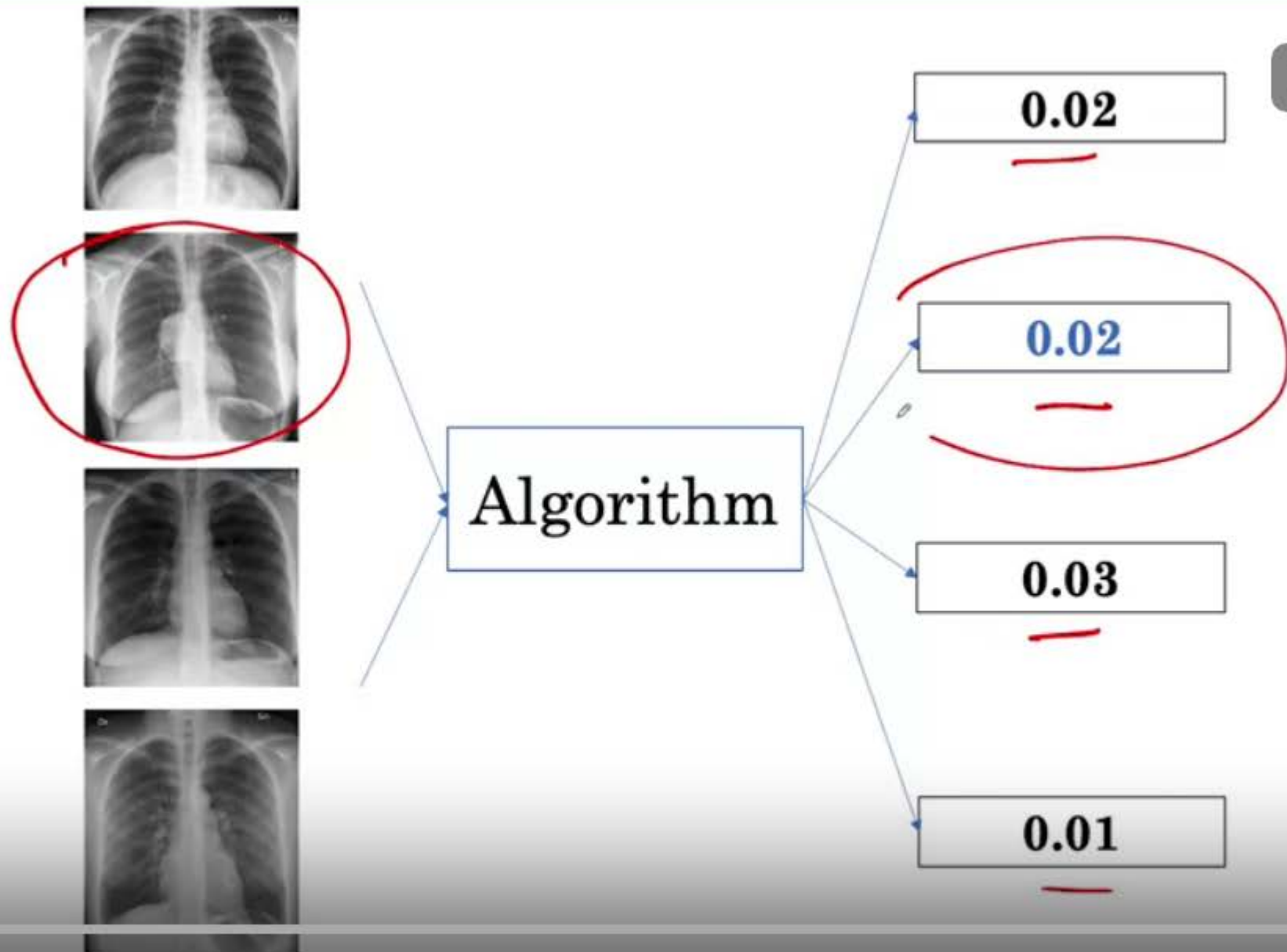
Mass

Normal

Normal

1.00

[Share](#)



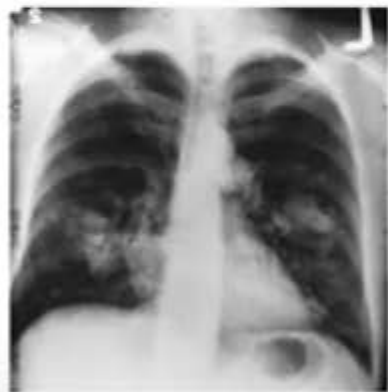
0:18 / 3:01



## Binary cross-entropy loss

$$\underline{L(X, y)} = \begin{cases} -\log P(Y = 1|X) & \text{if } \underline{y = 1} \\ -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

0 → 1



Algorithm

0.2

→  $P(Y=1|X)$

Label 1

$$L = -\log 0.2 \\ = \underline{0.70}$$



1:39 / 3:01





$$P(Y=0|X) = 1 - P(Y=1|X)$$

$$L(X, y) = \begin{cases} -\log P(Y = 1|X) & \text{if } y = 1 \\ -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$



Algorithm

0.7

Label 0

$$L = -\log(1 - 0.7) \\ = -\log(P(Y=0|X)) \\ = 0.52$$



3:00 / 3:01



1.00

$$L(X, y) = \begin{cases} -\log P(Y = 1|X) & \text{if } y = 1 \\ -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

[Share](#)

## Examples

## Prediction Probabilities

## Loss

P1 Normal	→	0.5	0.3
P2 Normal		0.5	0.3
P3 Normal		0.5	0.3
P4 Mass	→	0.5	0.3
P5 Normal		0.5	0.3
P6 Normal		0.5	0.3
P7 Mass		0.5	0.3
P8 Normal		0.5	0.3

$$-\log(1 - 0.5) = 0.3$$

$$-\log 0.5$$



0:46 / 3:18



$$L(X, y) = \begin{cases} -\log P(Y = 1|X) & \text{if } y = 1 \\ -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

## Examples

P1 Normal

P2 Normal

P3 Normal

P4 Mass

P5 Normal

P6 Normal

P7 Mass

P8 Normal

Loss

0.3

0.3

0.3

0.3

0.3

0.3

0.3

0.3

Total Loss From Mass Examples  $0.3 \times 2 = 0.6$ Total Loss From Normal Examples  $0.3 \times 6 = 1.8$

## Examples

P1 Normal

P2 Normal

P3 Normal

**P4 Mass**

P5 Normal

P6 Normal

**P7 Mass**

P8 Normal

$$L(X, y) = \begin{cases} -\log P(Y = 1|X) & \text{if } y = 1 \\ -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

$$L(X, y) = \begin{cases} \underline{w_p} \times -\log P(Y = 1|X) & \text{if } y = 1 \\ \underline{w_n} \times -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$



1.00

$$L(X, y) = \begin{cases} w_p \times -\log P(Y = 1|X) & \text{if } y = 1 \\ w_n \times -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

Share

## Examples

P1 Normal

P2 Normal

P3 Normal

P4 Mass

P5 Normal

P6 Normal

P7 Mass

P8 Normal

Loss

$$2/8 \times 0.3 = 0.075$$

$$2/8 \times 0.3 = 0.075$$

$$2/8 \times 0.3 = 0.075$$

Total Loss From Mass Examples ==  $0.225 \times 2 = 0.45$

Total Loss From Normal Examples =  $0.075 \times 6 = 0.45$

$$6/8 \times 0.3 = 0.225$$

$$2/8 \times 0.3 = 0.075$$

$$2/8 \times 0.3 = 0.075$$

$$6/8 \times 0.3 = 0.225$$

$$2/8 \times 0.3 = 0.075$$



2:30 / 3:18





## Examples

P1 Normal

P2 Normal

P3 Normal

**P4 Mass**

P5 Normal

P6 Normal

**P7 Mass**

P8 Normal

$$\underline{L(X, y)} = \begin{cases} \underline{w_p} \times -\log P(Y = 1|X) & \text{if } y = 1 \\ \underline{w_n} \times -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

$$\underline{w_p} = \frac{\text{num negative}}{\text{num total}}$$

$$\frac{6}{8}$$

$$\underline{w_n} = \frac{\text{num positive}}{\text{num total}}$$

$$\frac{2}{8}$$

Weighted Loss

3:17 / 3:18



## Examples

P1 Normal

P2 Normal

P3 Normal

P4 Mass

P5 Normal

P6 Normal

P7 Mass

P8 Normal

## Re-Sampled

P3 Normal

P6 Normal

P1 Normal

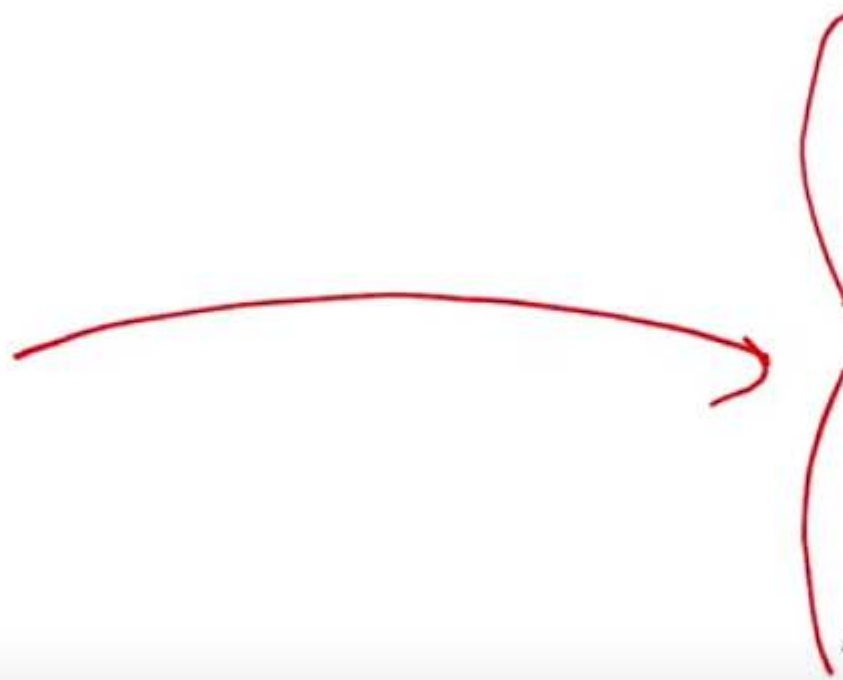
P8 Normal

P7 Mass

P4 Mass

P7 Mass

P4 Mass



## Examples

P1 Normal

P2 Normal

P3 Normal

P4 Mass

P5 Normal

P6 Normal

P7 Mass

P8 Normal

Normal  
P1, P2, P3, P5,  
P6, P8

Mass  
P4, P7

Sample 4

Sample 4

## Re-Sampled

P3 Normal

P6 Normal

P1 Normal

P8 Normal

P7 Mass

P4 Mass

P7 Mass

P4 Mass

Play



1:03 / 1:39



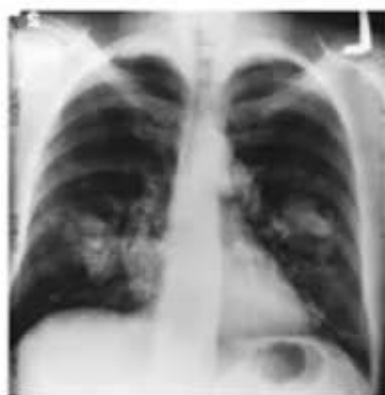
$$L(X, y) = \begin{cases} -\log P(Y = 1|X) & \text{if } y = 1 \\ -\log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

<u>Re-Sampled</u>	Prediction Probabilities	Loss
P3 Normal	0.5	0.3
P6 Normal	0.5	0.3
P1 Normal	0.5	0.3
P8 Normal	0.5	0.3
P7 Mass	0.5	0.3
P4 Mass	0.5	0.3
P7 Mass	0.5	0.3
P4 Mass	0.5	0.3

Total Loss From Mass Examples =  $0.3 \times 4 = 1.2$

Total Loss From Normal Examples =  $0.3 \times 4 = 1.2$

Re-sampling methods  
(Undersampling, Oversampling)



Algorithm

Mass?

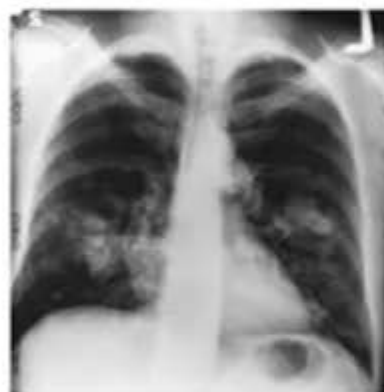
### 3 Key Challenges

Class Imbalance

**Multi-Task**

Dataset Size

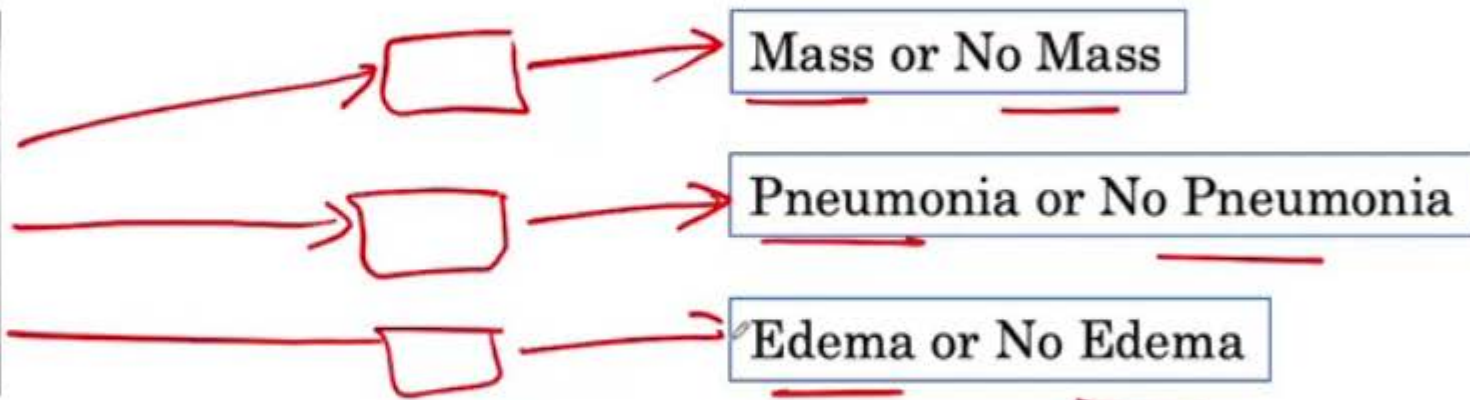
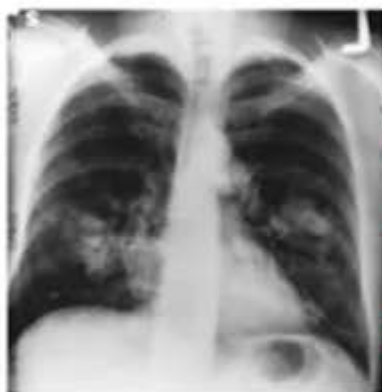


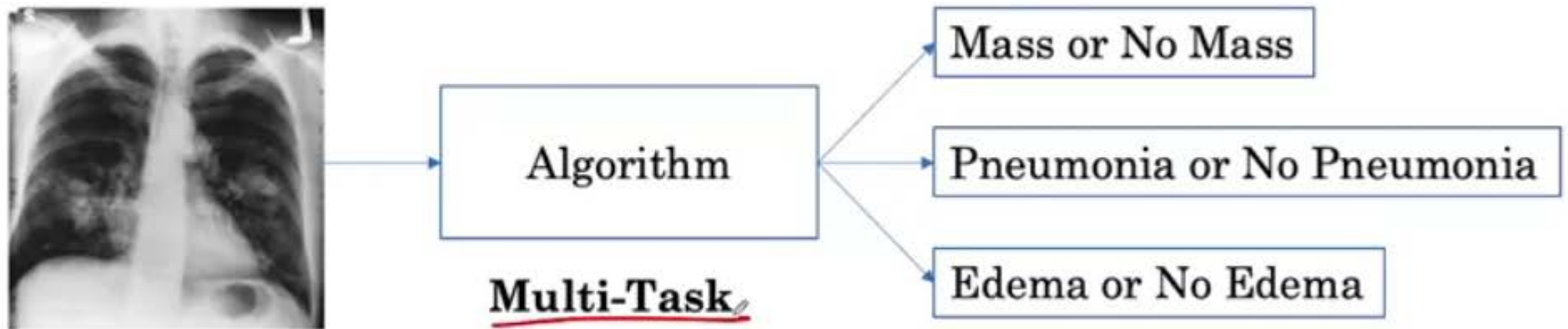


Algorithm

Mass or No Mass

Binary Task





Examples(mass, pneumonia, edema)

P1 0, 1, 0

P2 0, 0, 1

P3 0, 1, 1

P4 1, 0, 1

P5 1, 1, 1

P6 1, 0, 0

P7 0, 1, 1

P8 0, 0, 0

Prediction Probabilities

0.3, 0.1, 0.8

0.1, 0.1, 0.8

0.2, 0.2, 0.7

0.6, 0.3, 0.8

0.7, 0.7, 0.9

0.8, 0.1, 0.2

0.3, 0.9, 0.8

0.1, 0.1, 0.2

$$L(X, y)$$

?

## Examples

(mass, pneumonia, edema)

P1 0, 1, 0

P2 0, 0, 1

P3 0, 1, 1

P4 1, 0, 1

P5 1, 1, 1

P6 1, 0, 0

P7 0, 1, 1

P8 0, 0, 0

## Prediction Probabilities

0.3, 0.1, 0.8

0.1, 0.1, 0.8

0.2, 0.2, 0.7

0.6, 0.3, 0.8

0.7, 0.7, 0.9

0.8, 0.1, 0.2

0.3, 0.9, 0.8

0.1, 0.1, 0.2

$$\underline{L(X, y_{\text{mass}})} + \underline{L(X, y_{\text{pneumonia}})} + \underline{L(X, y_{\text{edema}})}$$

$$L(X, y)$$

Multi-Label / Multi-Task  
Loss



$$\underbrace{L(X, y_{\text{mass}})} + \underbrace{L(X, y_{\text{pneumonia}})} + \underbrace{L(X, y_{\text{edema}})}$$



### Examples

(mass, pneumonia, edema)

P1 0, 1, 0

P2 0, 0, 1

P3 0, 1, 1

P4 1, 0, 1

P5 1, 1, 1

P6 1, 0, 0

P7 0, 1, 1

P8 0, 0, 0

### Prediction Probabilities

0.3, 0.1, 0.8

0.1, 0.1, 0.8

0.2, 0.2, 0.7

0.6, 0.3, 0.8

0.7, 0.7, 0.9

0.8, 0.1, 0.2

0.3, 0.9, 0.8

0.1, 0.1, 0.2

### Loss

0.52 + 1.00 + 0.70

0.05 + 0.05 + 0.10

0.10 + 0.70 + 0.15

0.22 + 0.52 + 0.10

0.15 + 0.15 + 0.05

0.10 + 0.05 + 0.10

0.52 + 0.05 + 0.10

0.05 + 0.05 + 0.10



**Multi-Task**

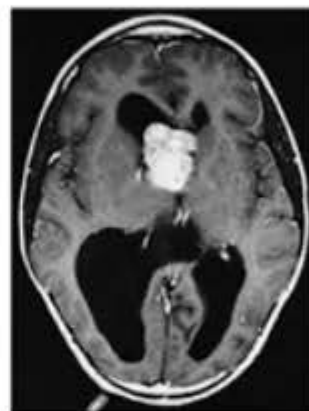
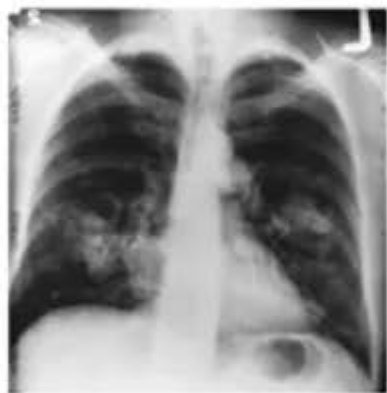
$$\underbrace{L(X, y_{\text{mass}})} + \underbrace{L(X, y_{\text{pneumonia}})} + \underbrace{L(X, y_{\text{edema}})}$$

$$L(X, y_{\text{mass}}) = \begin{cases} -\underline{w}_{\text{p}, \text{mass}} \log P(Y = 1|X) & \text{if } y = 1 \\ -\underline{w}_{\text{n}, \text{mass}} \log P(Y = 0|X) & \text{if } y = 0 \end{cases}$$

1.00

 Share

## Convolutional Neural Network



2:02 / 2:29

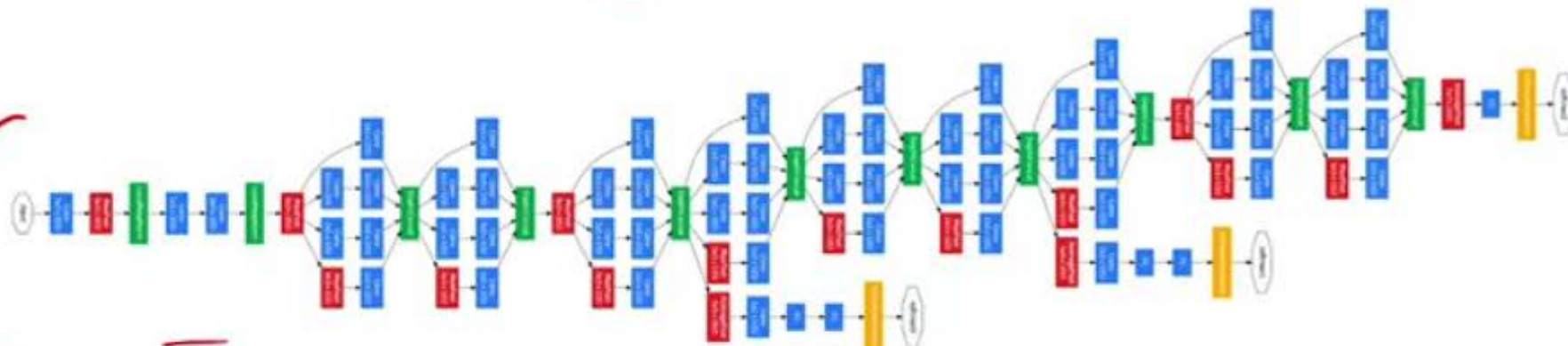


1.00

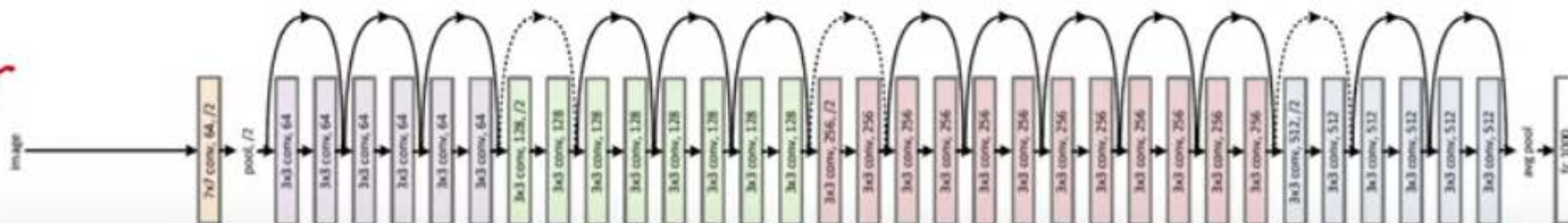
[Share](#)

## Convolutional Neural Network

### Inception-v3



### ResNet-34



### DenseNet

### ResNeXt

### EfficientNet



2:29 / 2:29

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1.00

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Millions of  
Images with  
Labels



~10k or ~100k  
Images with  
Labels



0:13 / 2:54

 deeplearning.ai





## 1. Pretraining



Convolutional  
Neural Network

Penguin or Cat or Dog

Copy / Transfer learned features

Convolutional  
Neural Network



0:39 / 2:54



## 1. Pretraining



Convolutional  
Neural Network

Penguin or Cat or Dog

## 2. Fine-tuning



Convolutional  
Neural Network

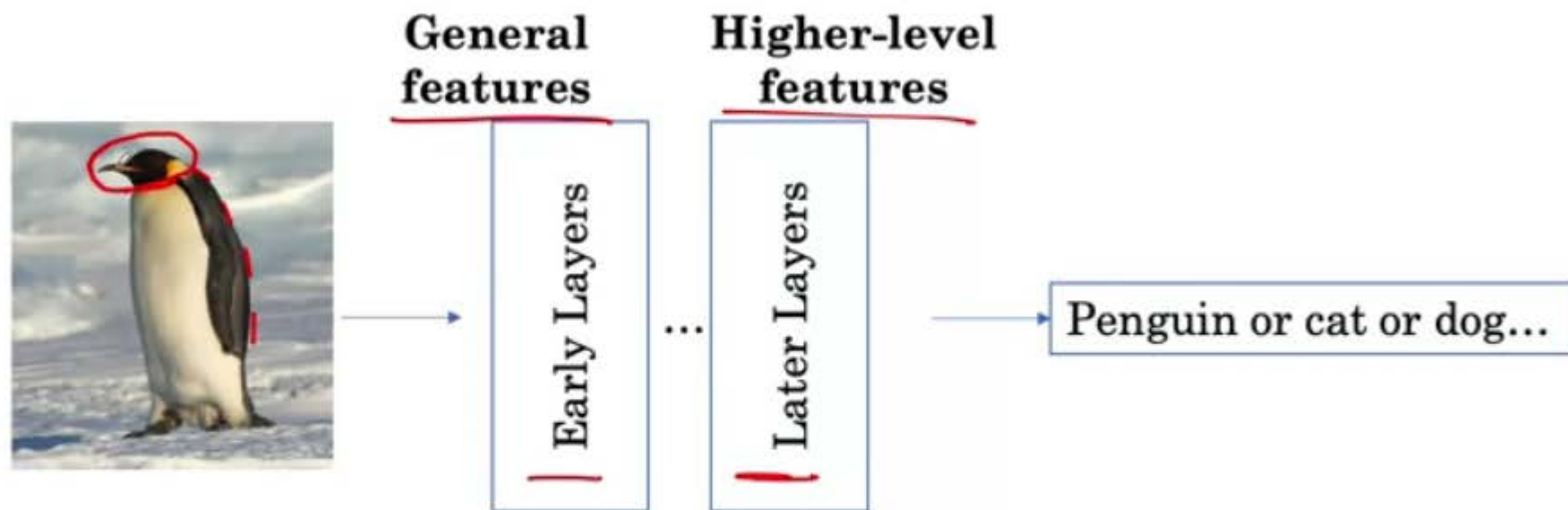
Update

Mass or No Mass

Pneumonia or No Pneumonia

Edema or No Edema

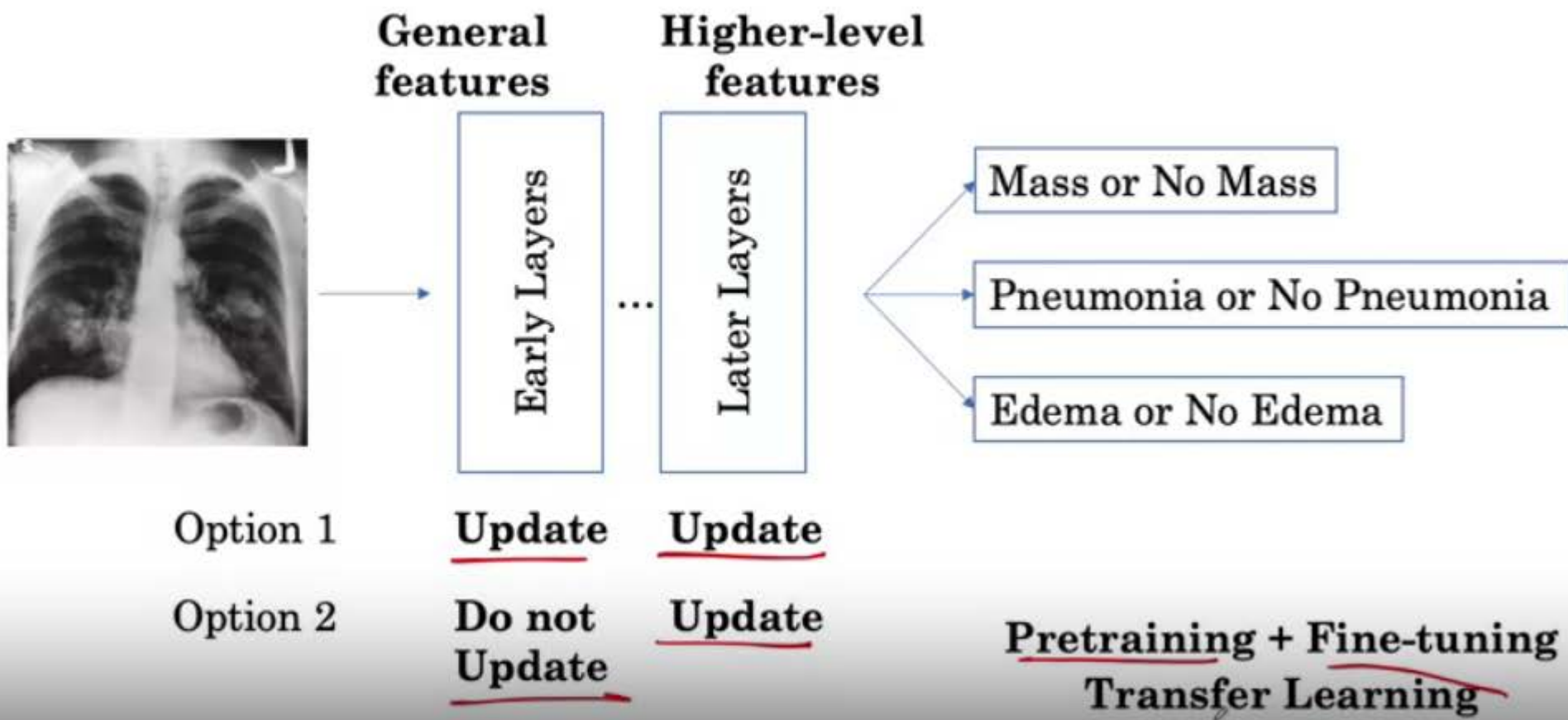




2:12 / 2:54



## 2. Fine-tuning



Play



2:51 / 2:54





**Mass**



**Mass**



**Mass**



**Mass**

**Data Augmentation**



## Do Augmentations Reflect Variations In Real World?



**Mass**



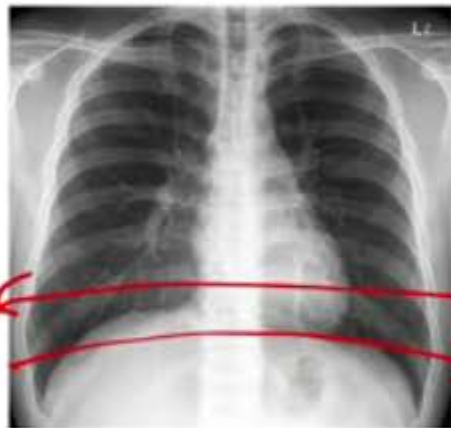
**Mass**



1:06 / 3:12



## Do Augmentations Keep the Label the Same?



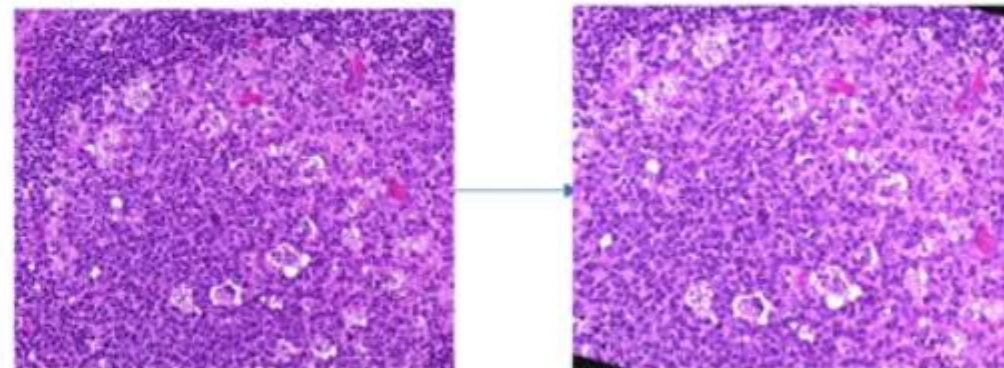
Normal



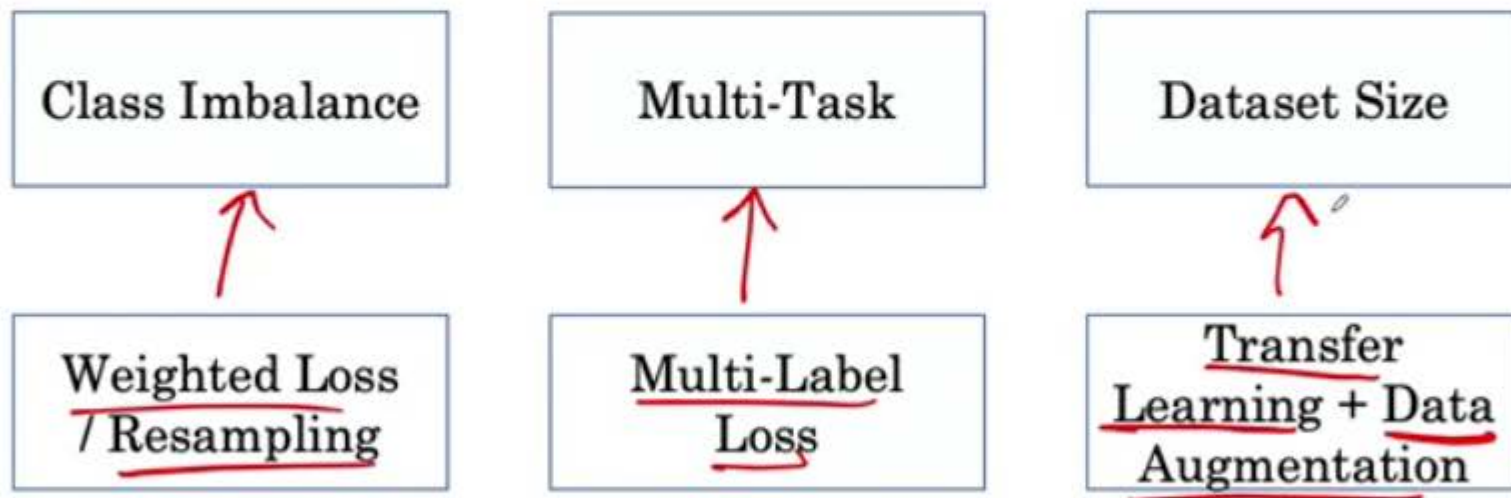
Normal  
Or  
Dextrocardia?

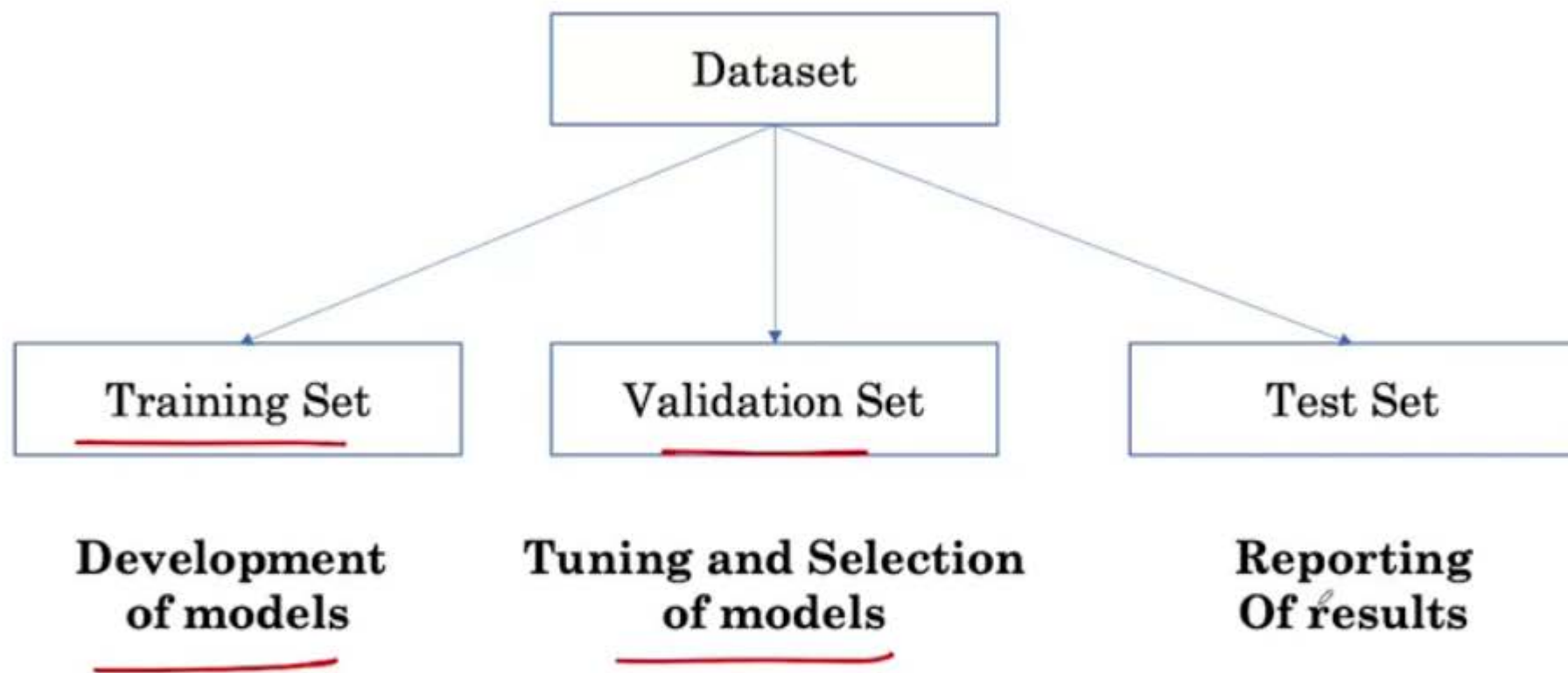


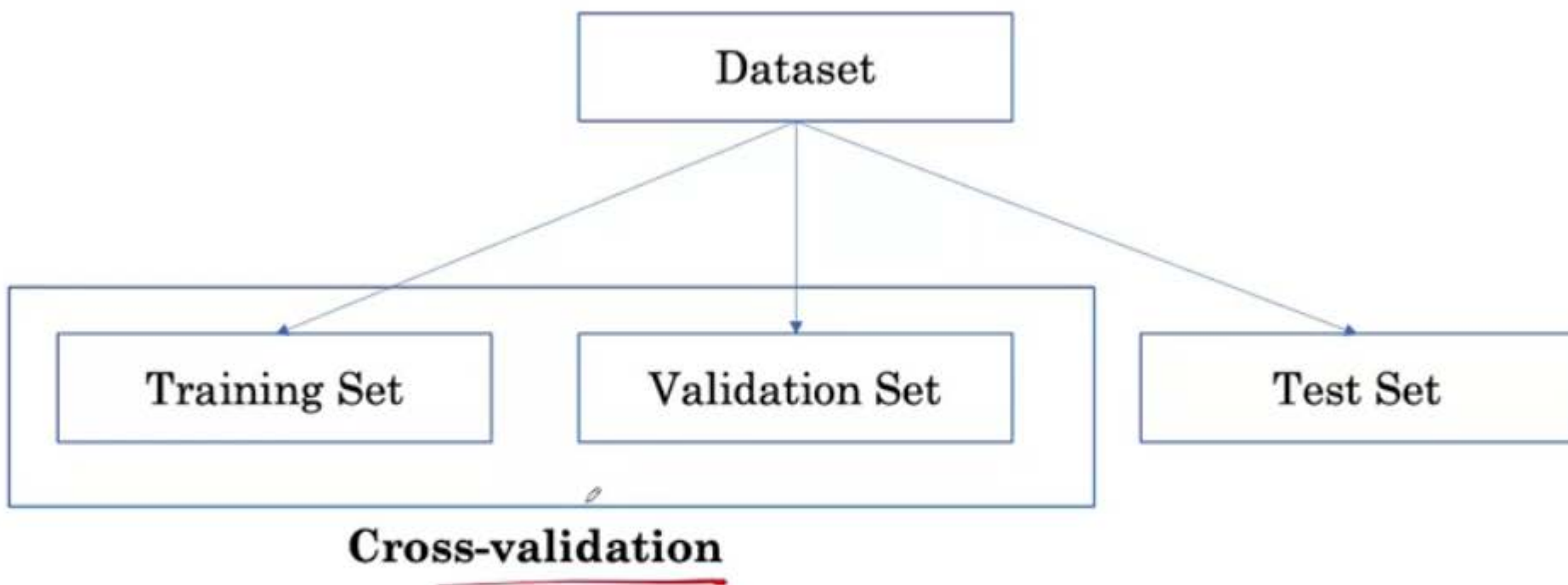
**Rotate + Flip**



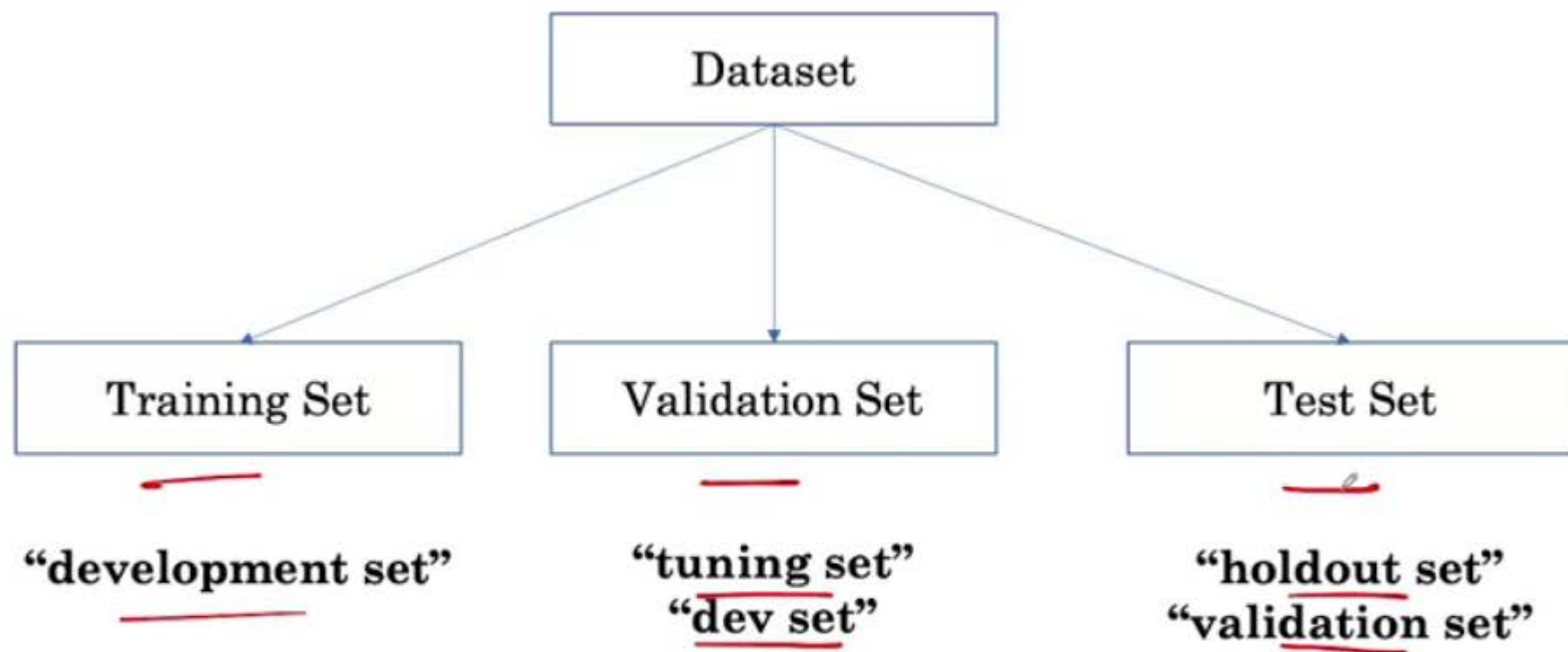
**Rotate + Crop + Color Noise**





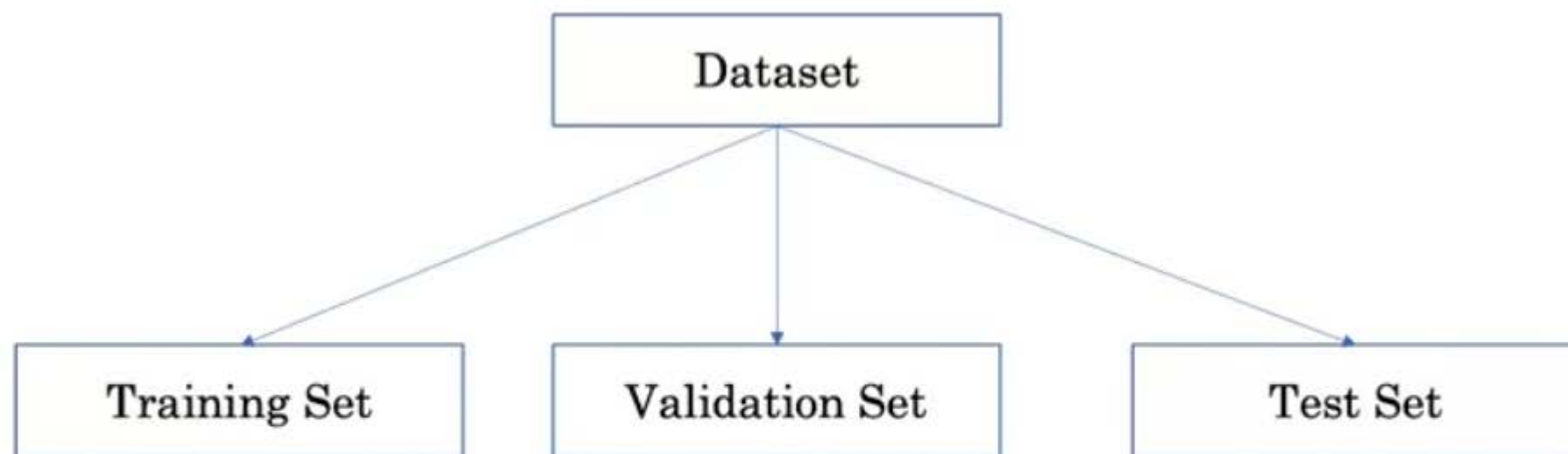






1.00

 Share



### 3 Key Challenges

**Patient Overlap**


**Set Sampling**

**Ground Truth**

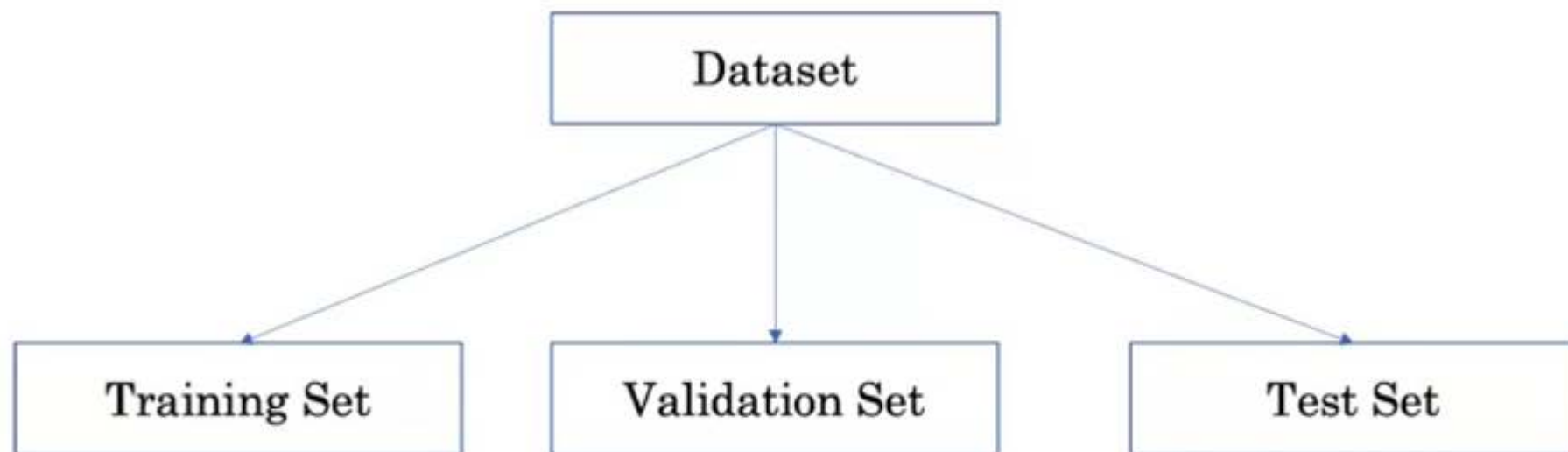
Play



1:32 / 2:49

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### 3 Key Challenges

Patient Overlap

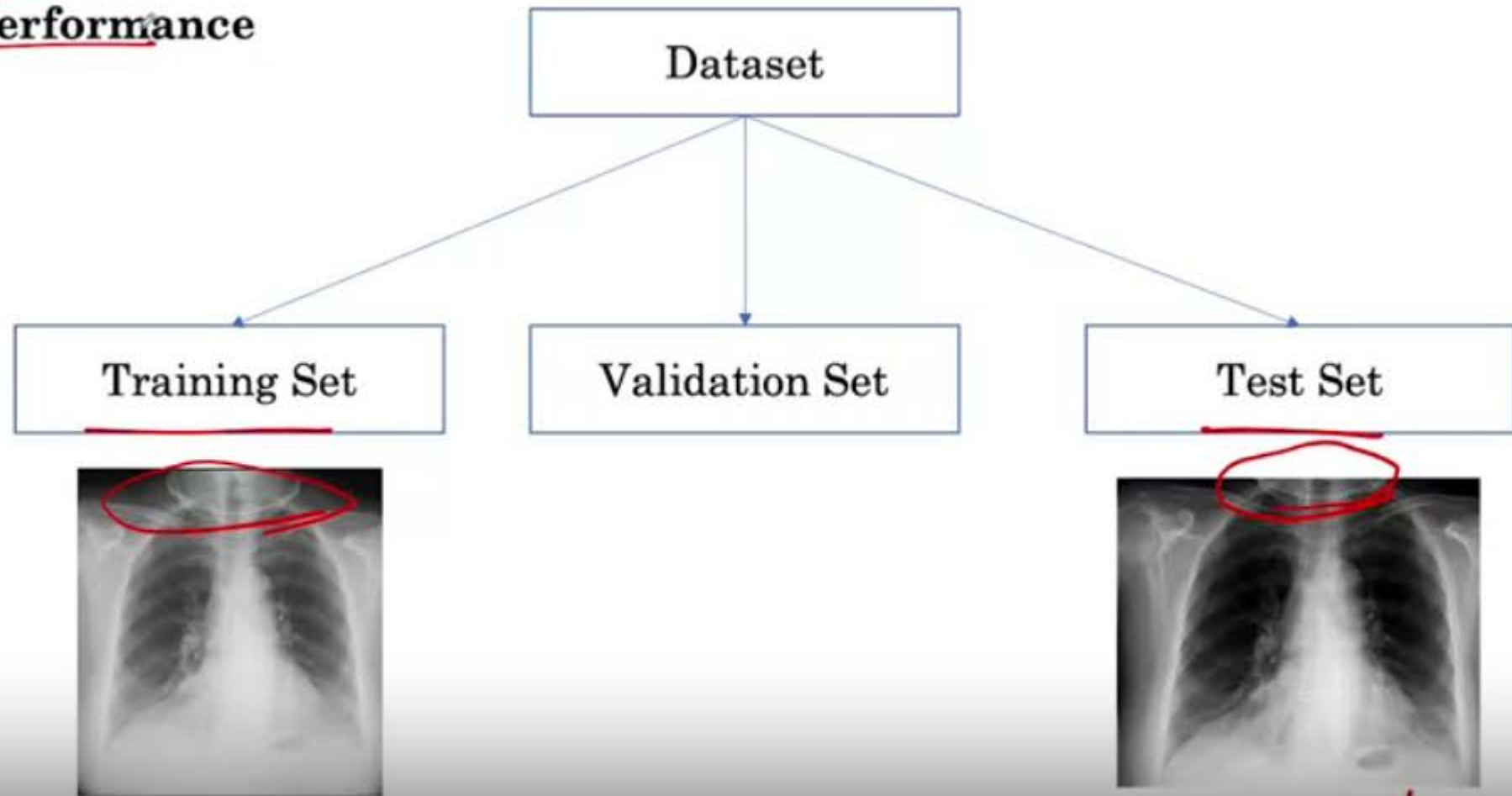
Set Sampling

Ground Truth

1.00

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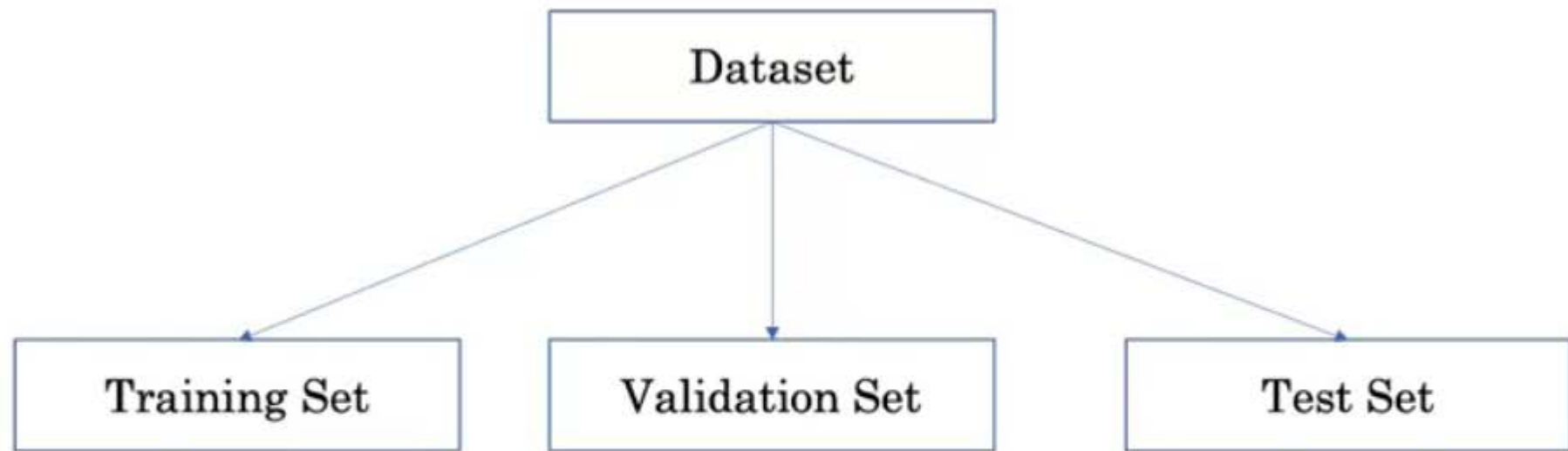
## Over-optimistic Test set Performance



2:46 / 2:49

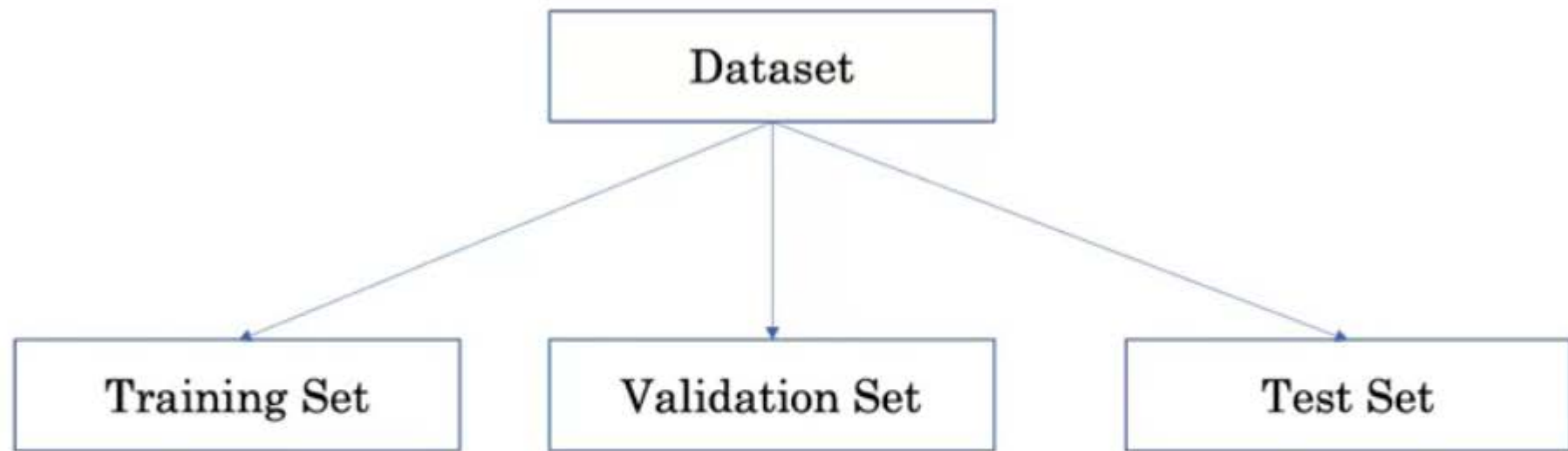
deeplearning.ai





0:22 / 1:36





0:22 / 1:36





## Split By Image

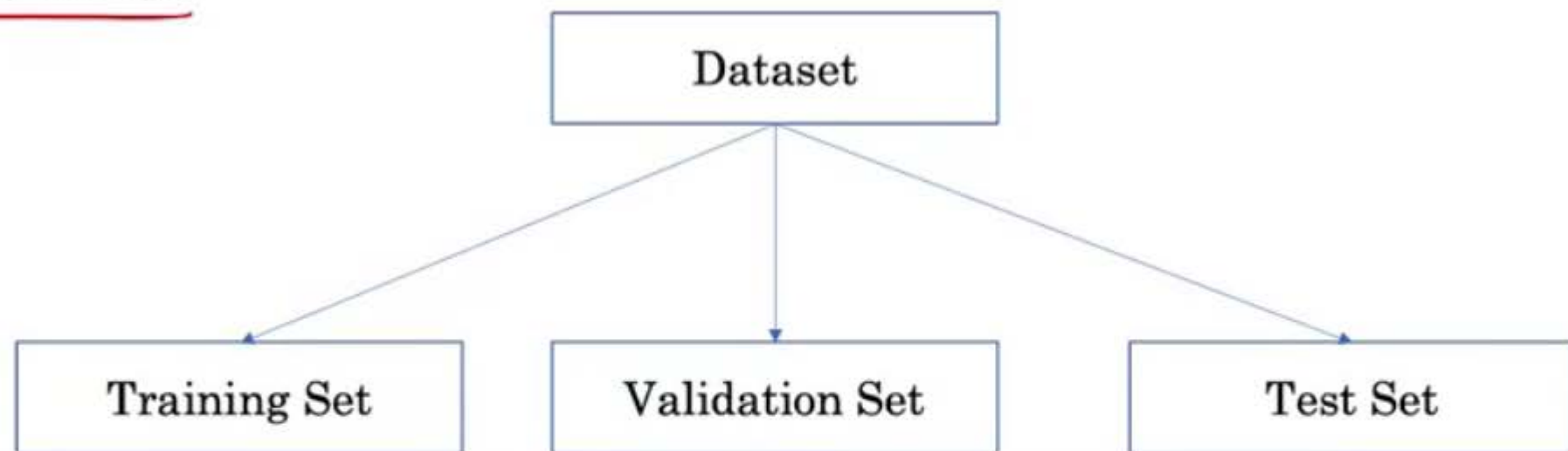


Image	Patient id	Label (Mass)	...	Image	Patient id	Label (Mass)	...	Image	Patient id	Label (Mass)	...
<u>xray1.jpg</u>	<u>20</u>	1		<u>xray2.jpg</u>	<u>20</u>	1		<u>xray0.jpg</u>	<u>20</u>	1	
xray4.jpg	17	0		xray5.jpg	11	0		xray3.jpg	38	0	
xray7.jpg	32	0		xray9.jpg	32	0		xray8.jpg	32	0	
...				...				...			

## Patient Overlap

## Split By Patient

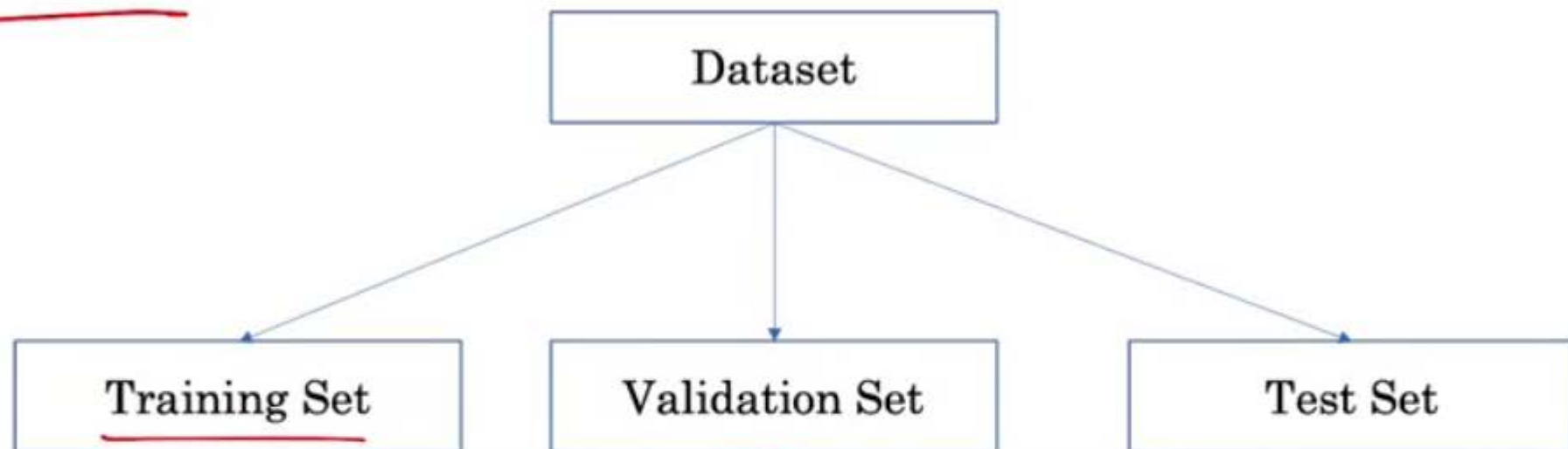


Image	Patient id	Label (Mass)	...	Image	Patient id	Label (Mass)	...	Image	Patient id	Label (Mass)	...
<u>xray1.jpg</u>	<u>20</u>	1		<u>xray7.jpg</u>	<u>32</u>	1		xray4.jpg	17	0	
<u>xray2.jpg</u>	<u>20</u>	1		<u>xray8.jpg</u>	<u>32</u>	0		xray3.jpg	38	0	
<u>xray0.jpg</u>	<u>20</u>	1		<u>xray9.jpg</u>	<u>32</u>	0		xray5.jpg	11	0	
...				...				...			

1.00

~10k or ~100k  
Patients with  
Labels

Image	Patient id	Label (Mass)	...
xray10.jpg	15	0	
xray23.jpg	24	0	
xray31.jpg	20	0	
xray41.jpg	56	1	
...			

Dataset

Test Set

10%

**Sizes in Studies**  
120 CT scans, 400-500 X-  
Rays, 130 Whole Slides

Image	Patient id	Label (Mass)	...
xray4.jpg	17	0	
xray3.jpg	38	0	
xray5.jpg	11	0	
...			

Image	Patient id	Label (Mass)	...
xray10.jpg	15	0	
xray23.jpg	24	0	
xray31.jpg	20	0	
xray41.jpg	56	1	
...			

Sample Test Set,  
then Validation,  
then Train

Dataset

Validation Set

Test Set

Sample to have same  
distribution of classes as  
the test set.

1.00

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Image	Patient id	Label (Mass)	...
xray10.jpg	15	0	
xray23.jpg	24	0	
xray31.jpg	20	0	
xray41.jpg	56	1	
...			

Sample Test Set,  
then Validation,  
then Train

Dataset

Remaining Patients  
in Training Set.

Training Set

Validation Set

Test Set



2:47 / 2:49



1.00

 Share

Ground Truth /  
Reference Standard

Play



0:12 / 1:31

 deeplearning.ai





1.00

 Share



Pneumonia

Other Disease

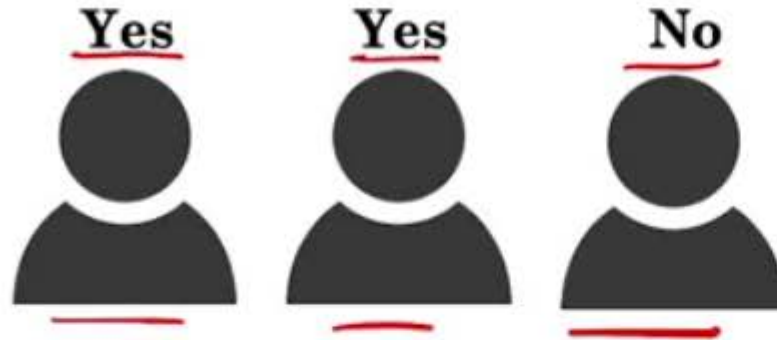


inter-observer disagreement



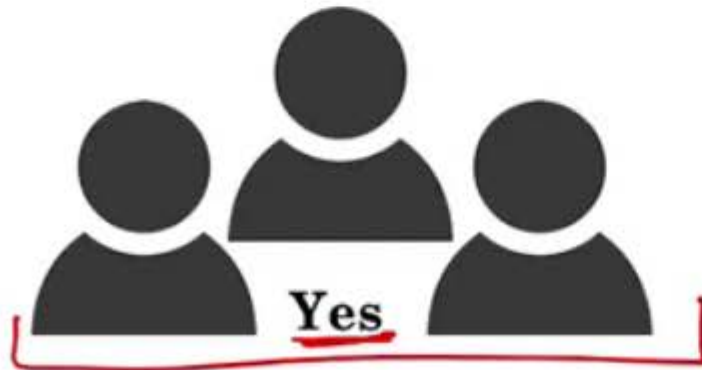
0:42 / 1:31





Is it pneumonia? Yes

### Consensus Voting



Is it pneumonia? Yes



1.00

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## CT Confirmation



Mass

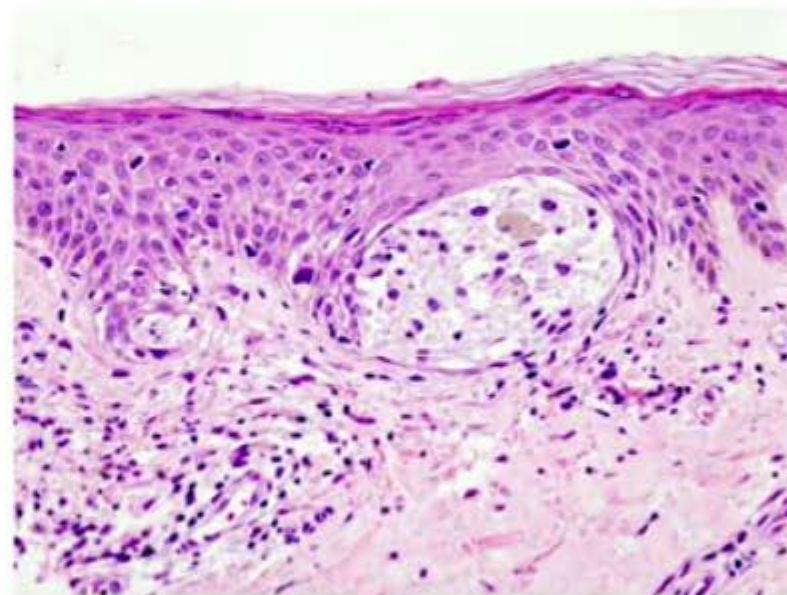
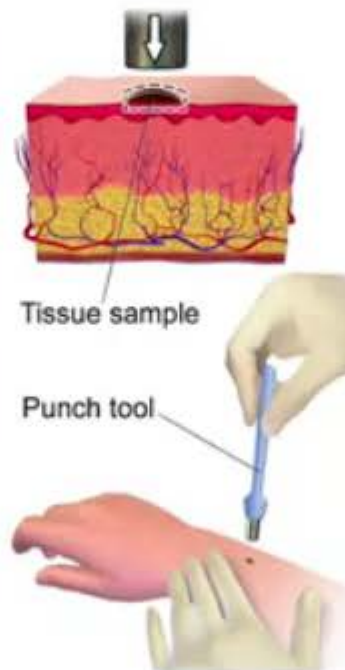


0:27 / 2:32



1.00

## Skin Biopsy



Play

Cancerous

1.00

 Share

Ground Truth  
Reference Standard

**Consensus voting**

**More definitive test**



1:29 / 2:32



### 3 Key Challenges

Patient Overlap  
**Split by Patient**

Set Sampling  
**Minority class**  
**Sampling**

Ground Truth  
**Consensus**  
**voting / more**  
**definitive test**

