

Détection et diagnostic de tumeurs sur échographie mammaire

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Echographie pour le cancer du sein

Dépistage

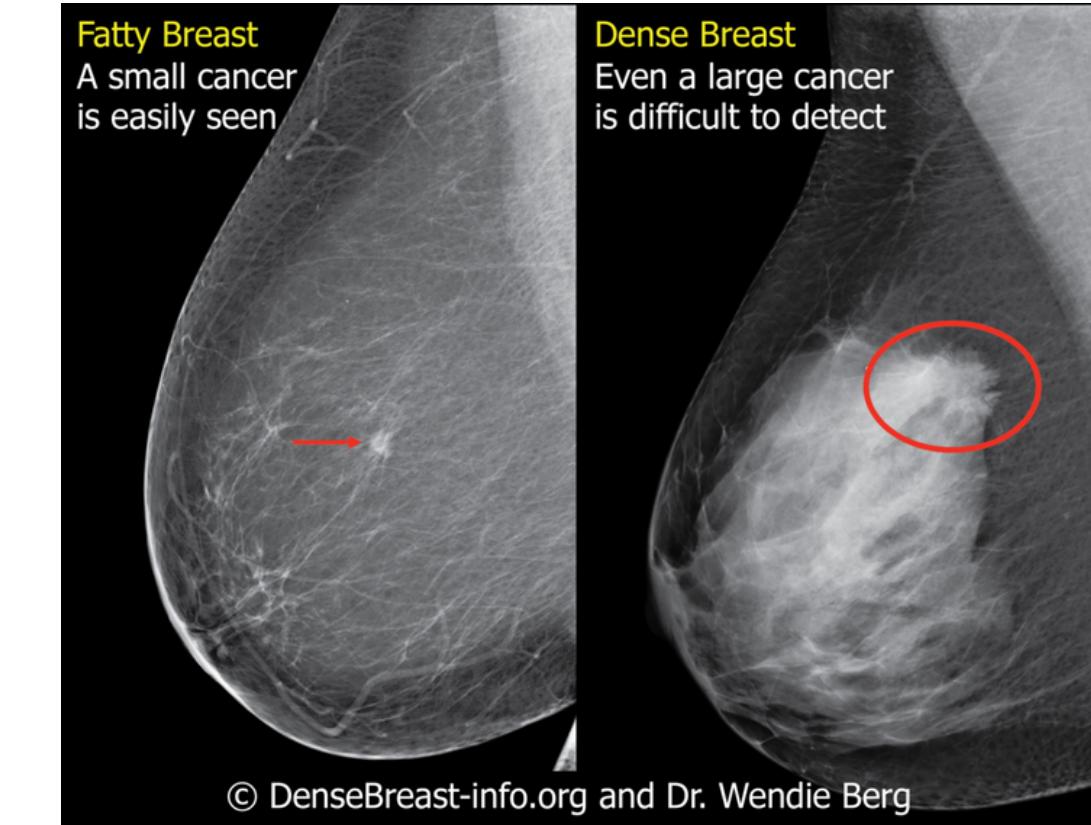
Examen de référence = mammographie

Examens complémentaires possibles

Echographie :

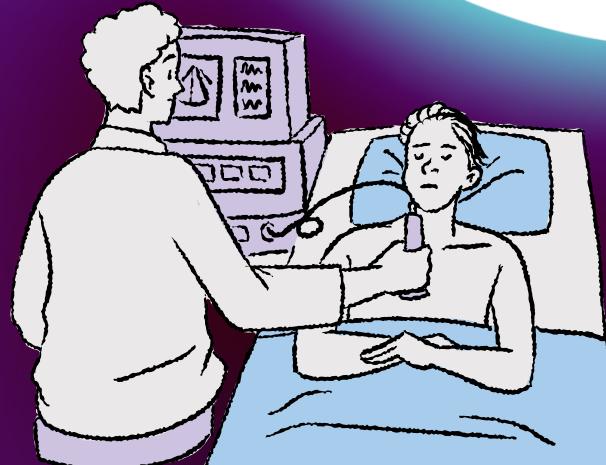
- en cas de **seins dense**, pour détecter des lésions occultes sur la mammographie
- pour **analyser** finement une anomalie détectée sur la mammographie

Pratiquée par un médecin radiologue ou un manipulateur radio



Diagnostic

Prescription d'échographie mammaire en présence d'une masse suspecte pour l'analyser



Présentation du dataset

Sujets :

600 femmes âgées de 25 à 75 ans

Données :

780 images d'**échographie mammaire** avec le **masque de segmentation** correspondant

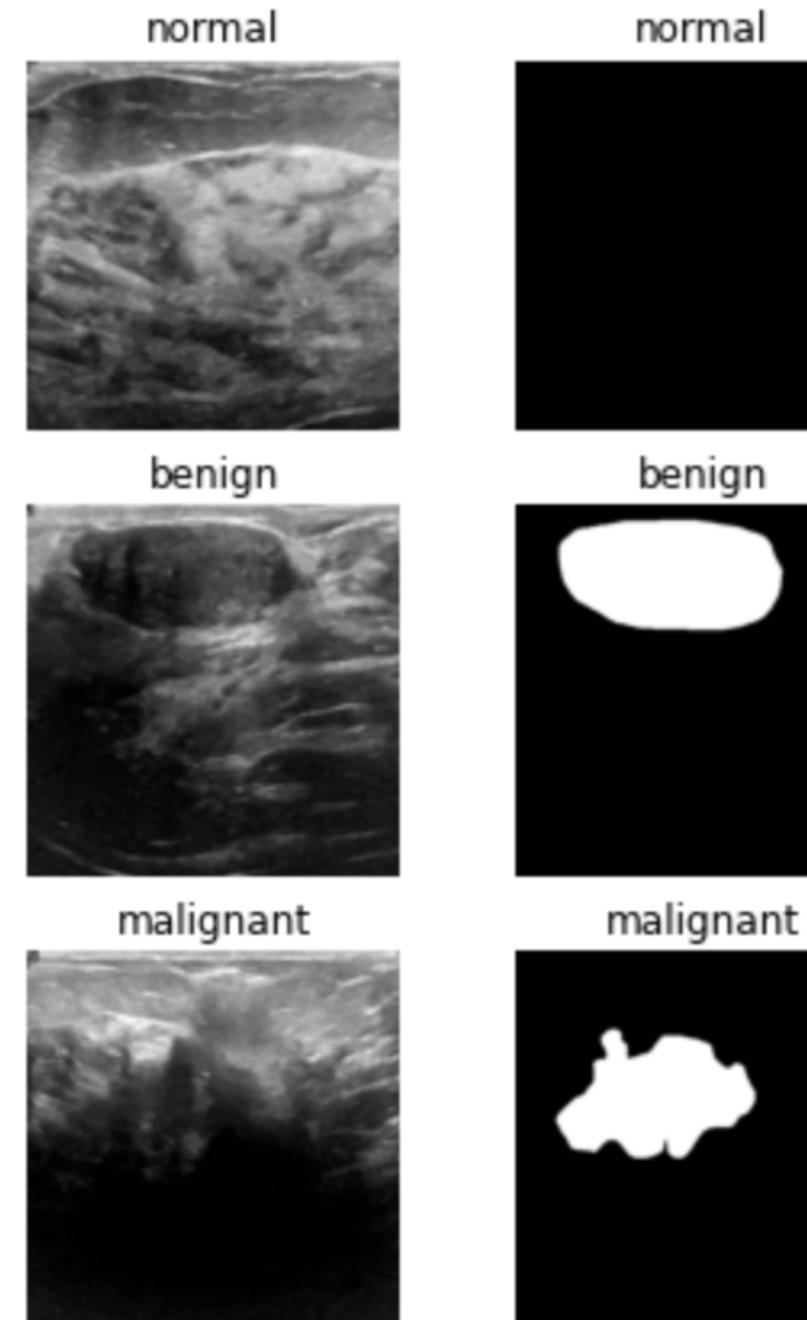
Taille d'image moyenne de **500*500** pixels

Image au format PNG

Classification :

Images classées en trois catégories :
normales, bénignes et **malignes**.

Présentation du dataset



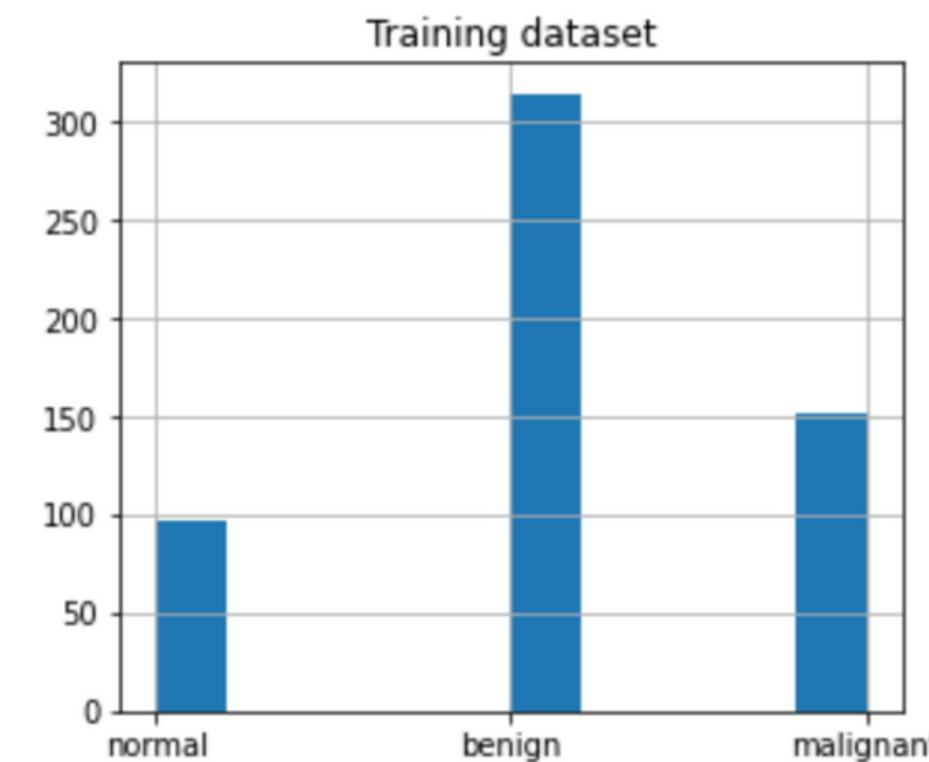
Examen **normal**

Tumeur **bénigne**

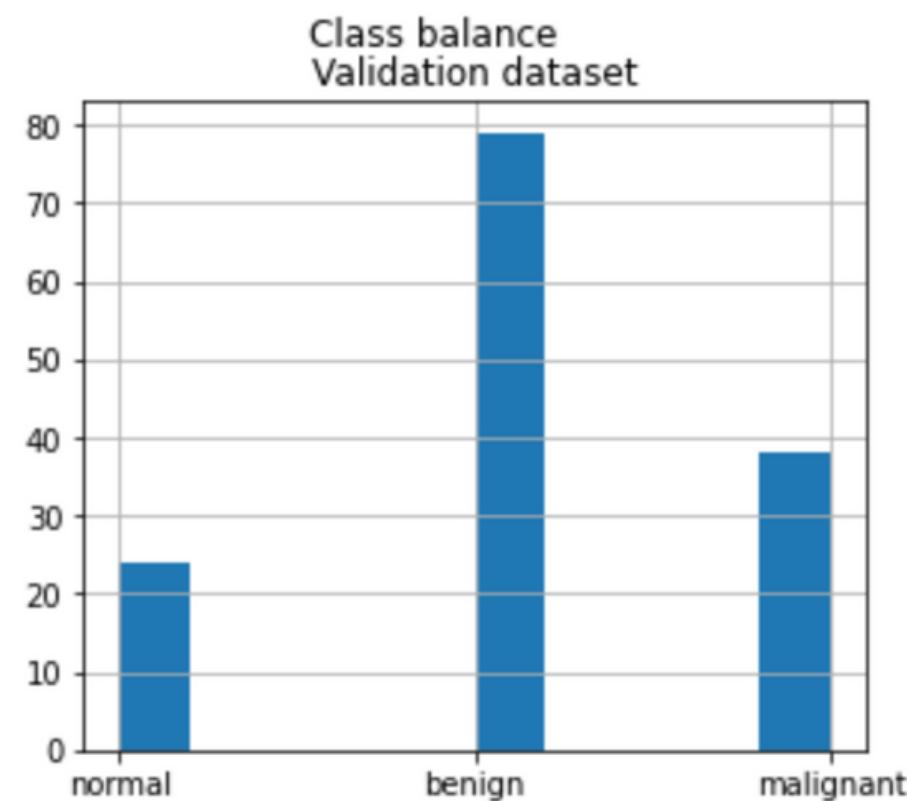
Tumeur **maligne**

2 tâches : Segmentation + Classification

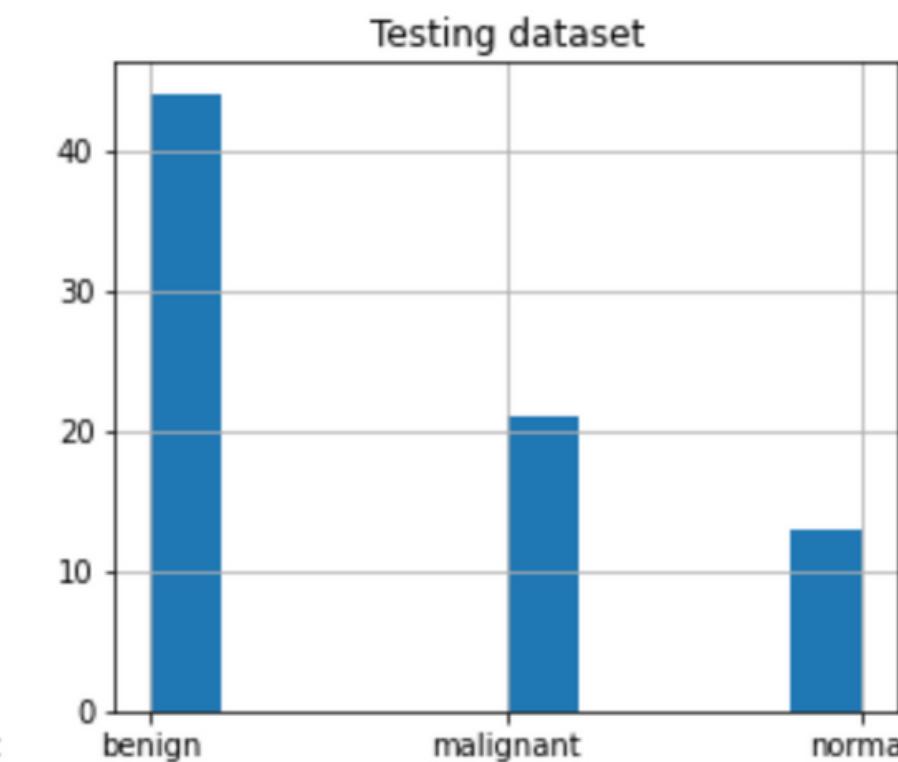
Split du dataset



561 samples



141 samples

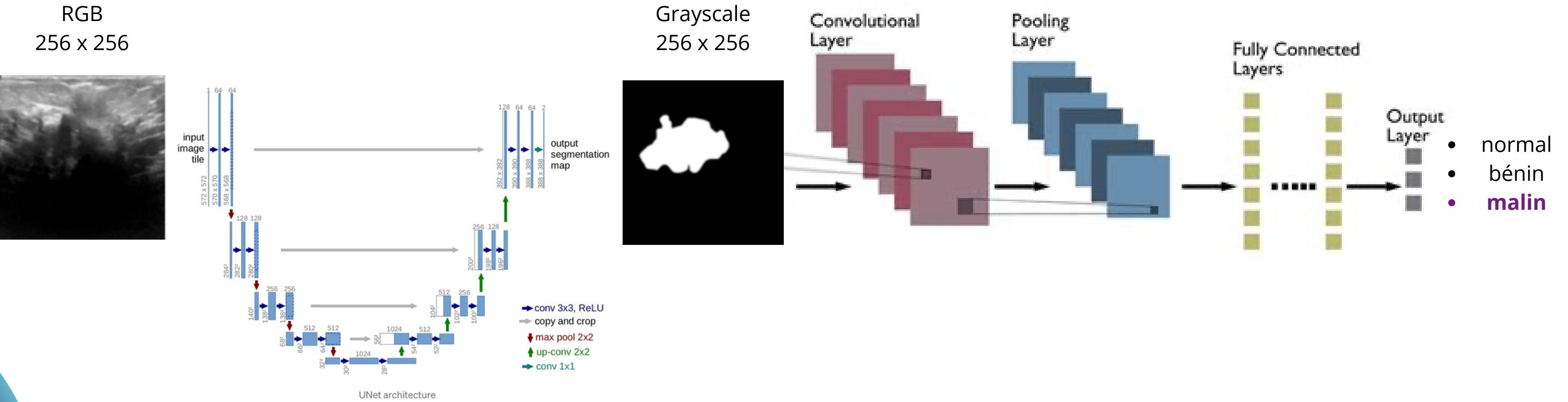


78 samples

02

Modèles séquentiels

Architecture en deux étapes



Unet
encoder = resnet34
pre-trained sur ImageNet

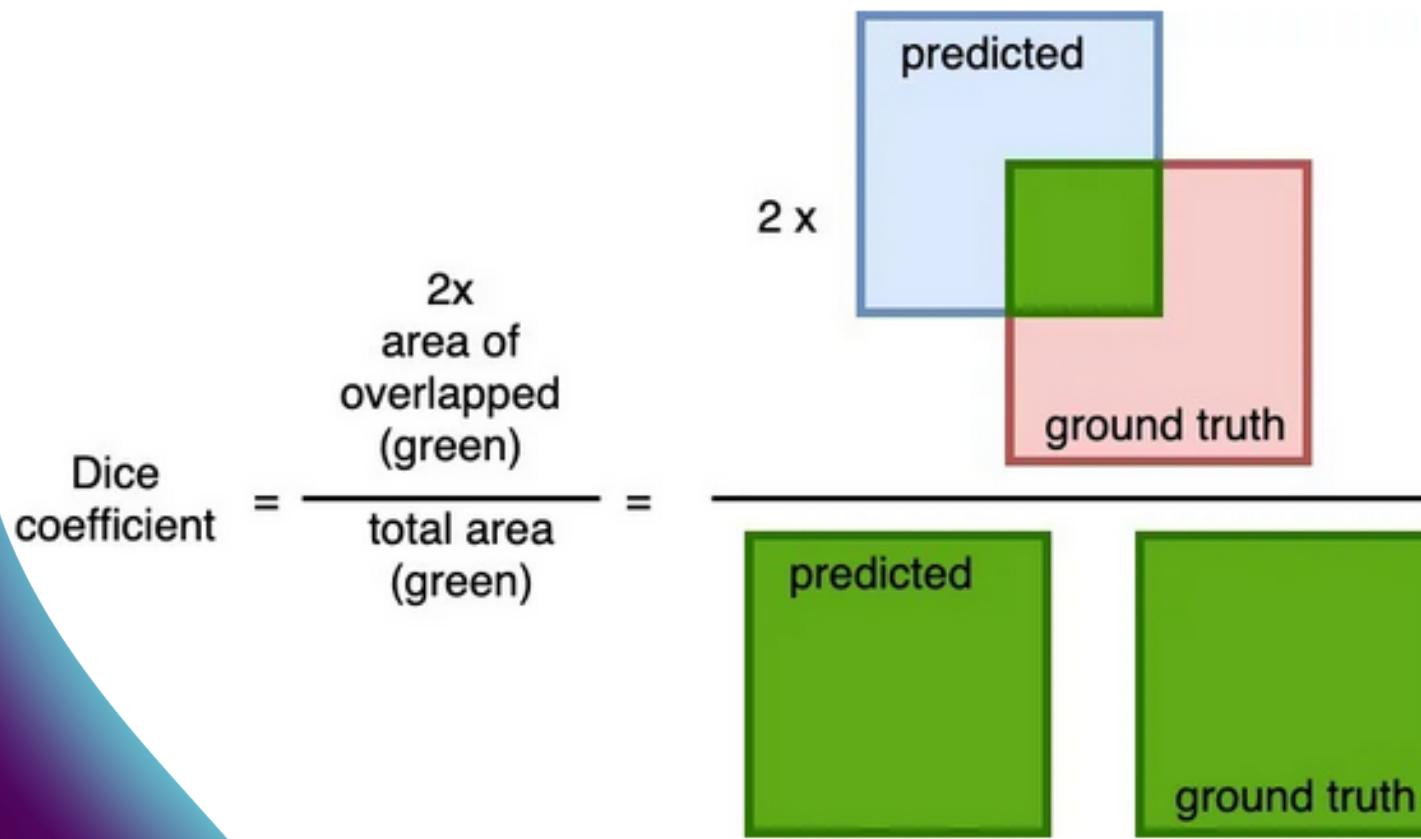
Resnet50
pre-trained sur ImageNet

Loss

Segmentation

Mixed Loss = Focal + Dice Loss

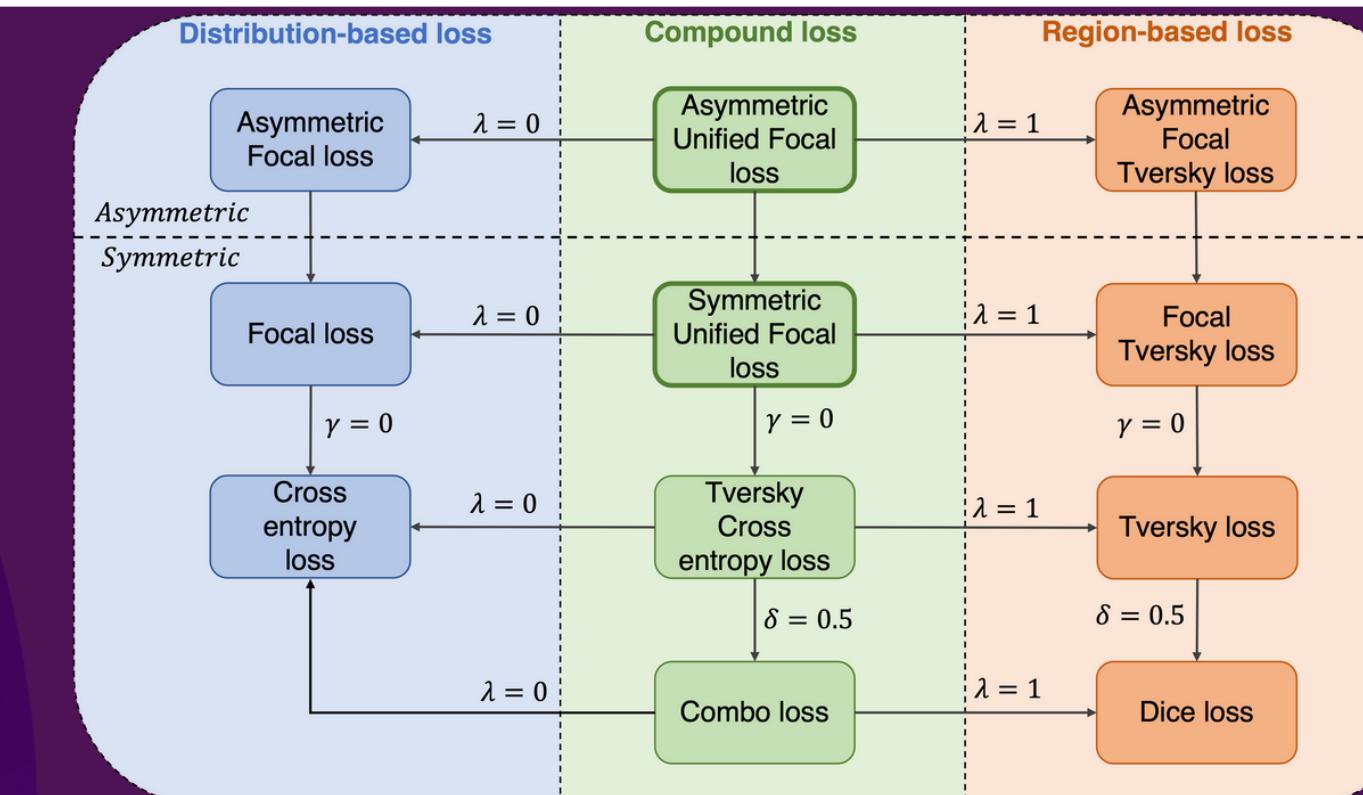
$$FocalLoss = - \sum_i (1 - P(i))^\gamma \log P(i)$$



Classification

Cross Entropy Loss

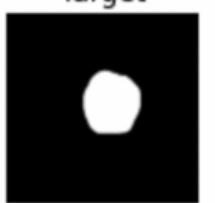
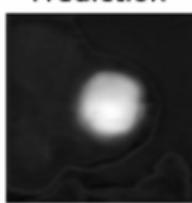
$$H(P^*|P) = - \sum_i P^*(i) \log P(i)$$



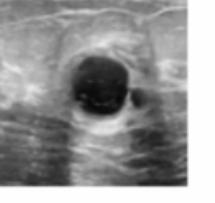
Résultats individuels - mask



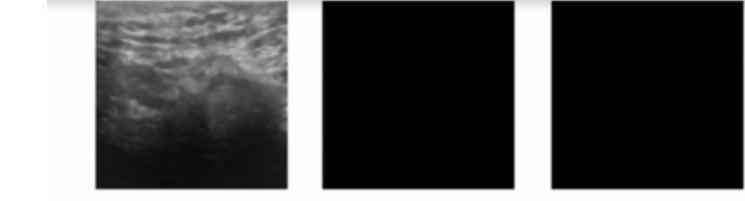
Ultrasound Image Target Prediction

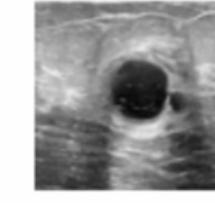





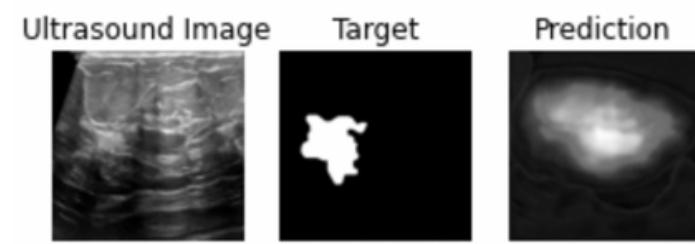
Ultrasound Image Target Prediction




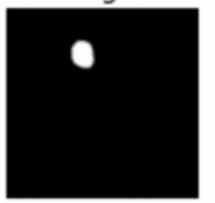
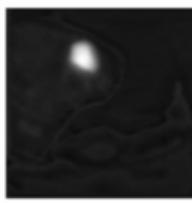


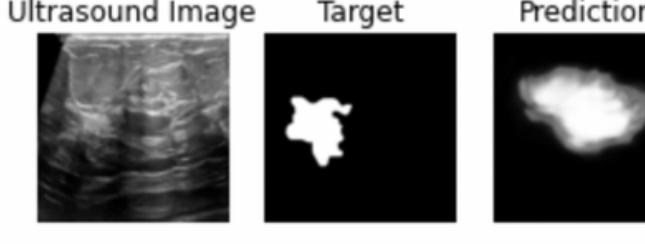
Ultrasound Image Target Prediction


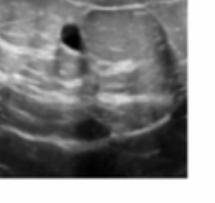


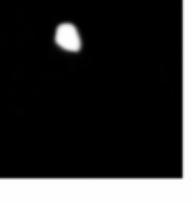


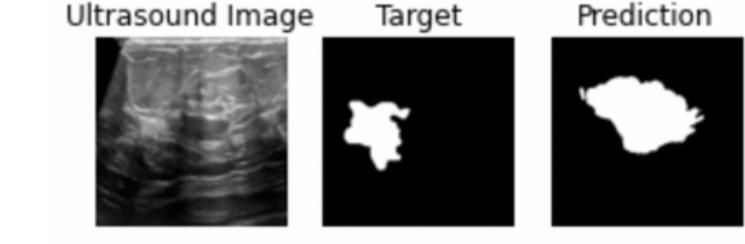
Ultrasound Image Target Prediction

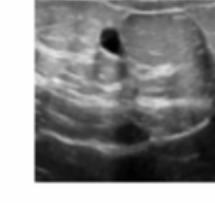





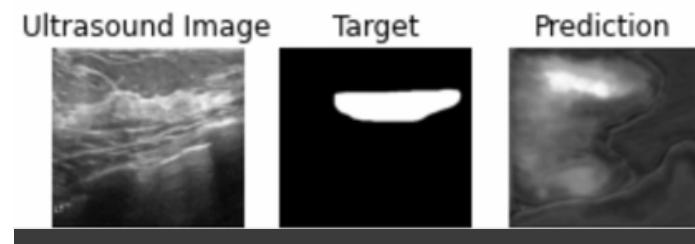
Ultrasound Image Target Prediction




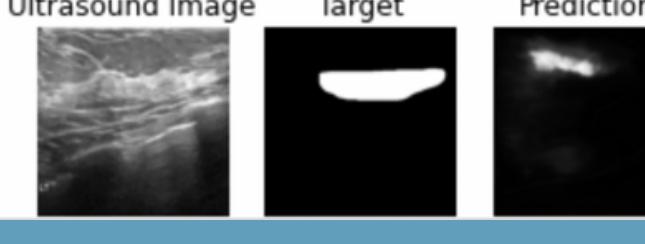


Ultrasound Image Target Prediction


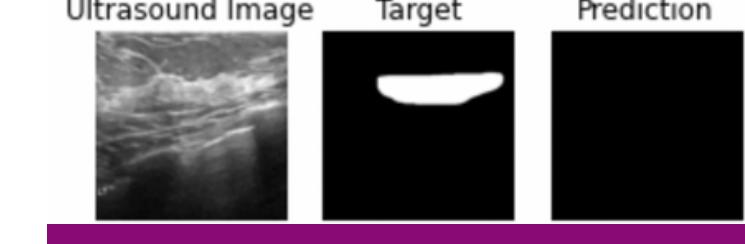




Output sans traitement

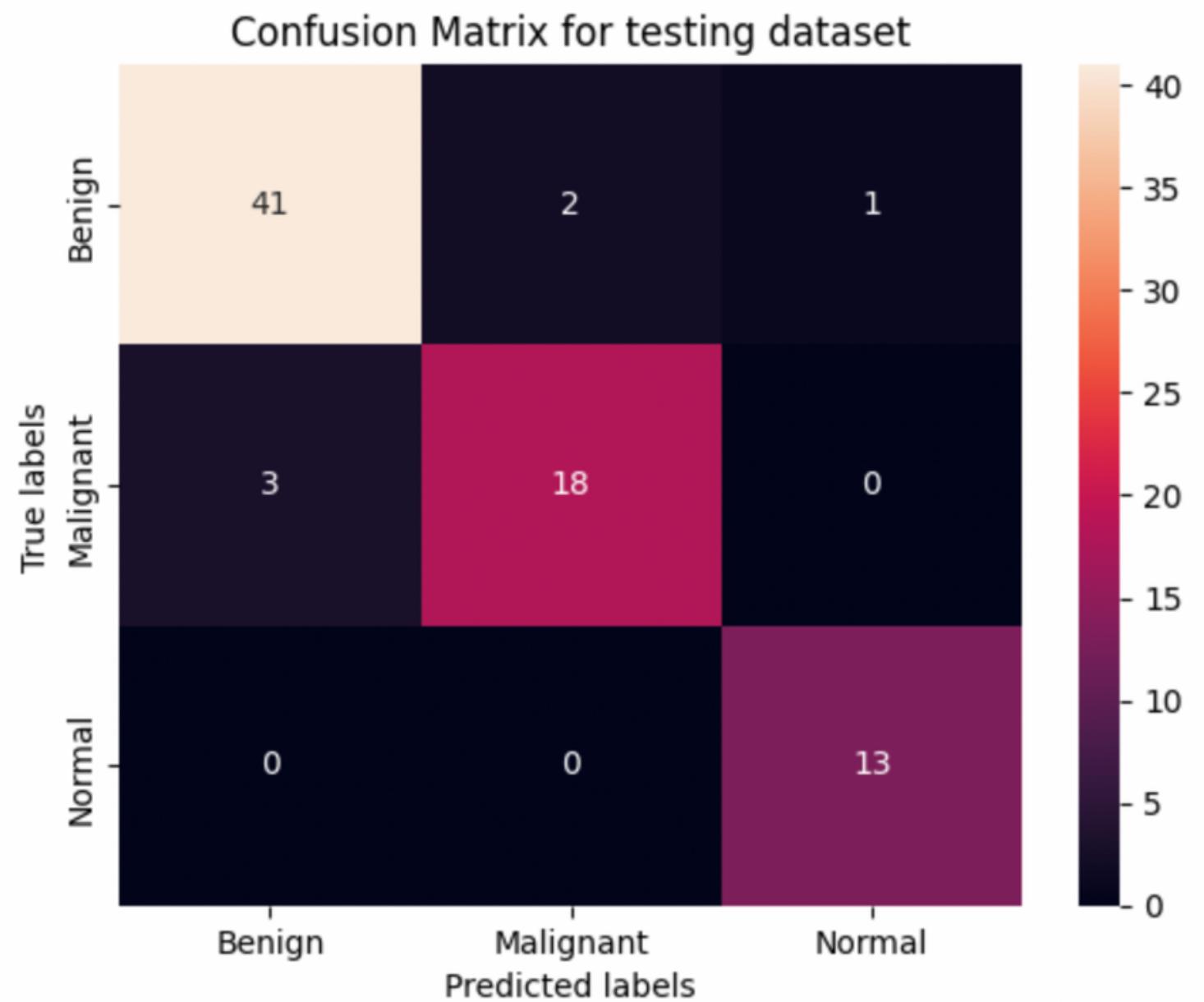


Application d'une sigmoid
sur l'output



Application d'une sigmoid
et d'un arrondi sur l'output

Résultats individuels - labels



Accuracy Test Dataset :

Accuracy Validation Dataset : **98.6%**

Predictions



Normal



Malir



Bénin

Predictions



Malin



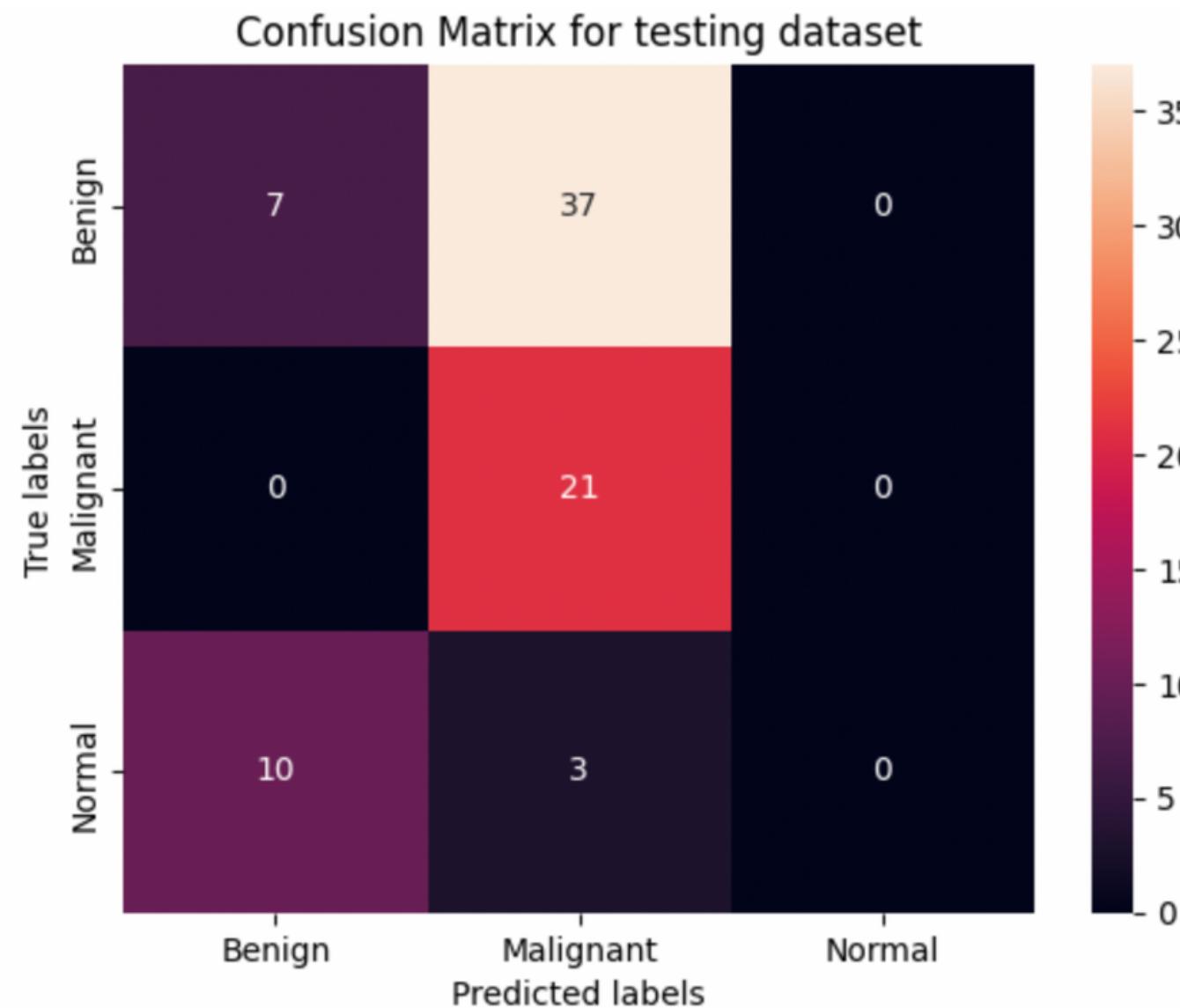
Bénin



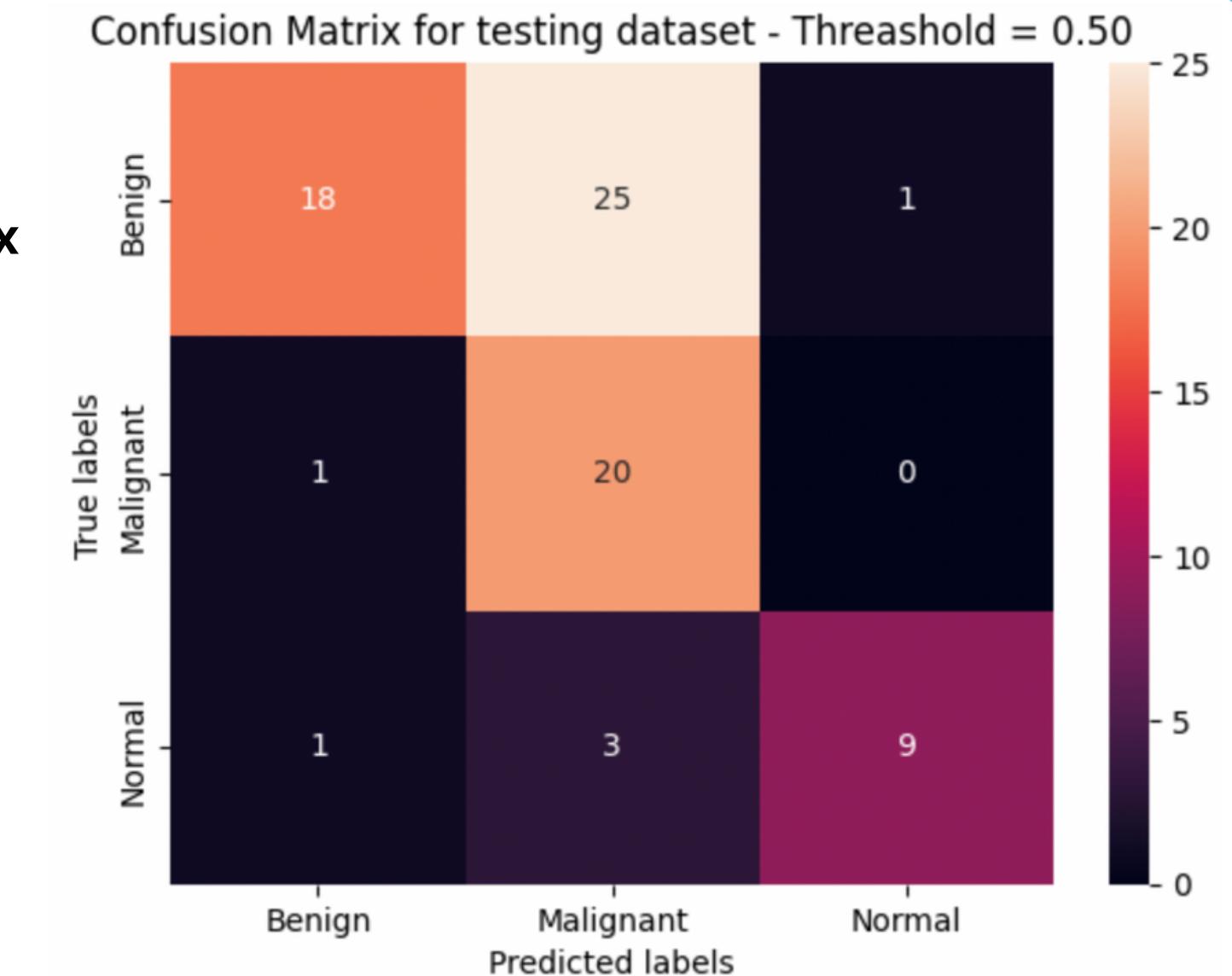
Malin

Résultats combinés - labels

Premiers résultats : aucune prédition des normaux et beaucoup d'erreurs



Adaptation des conditions : l'argmax ne semble pas pertinent

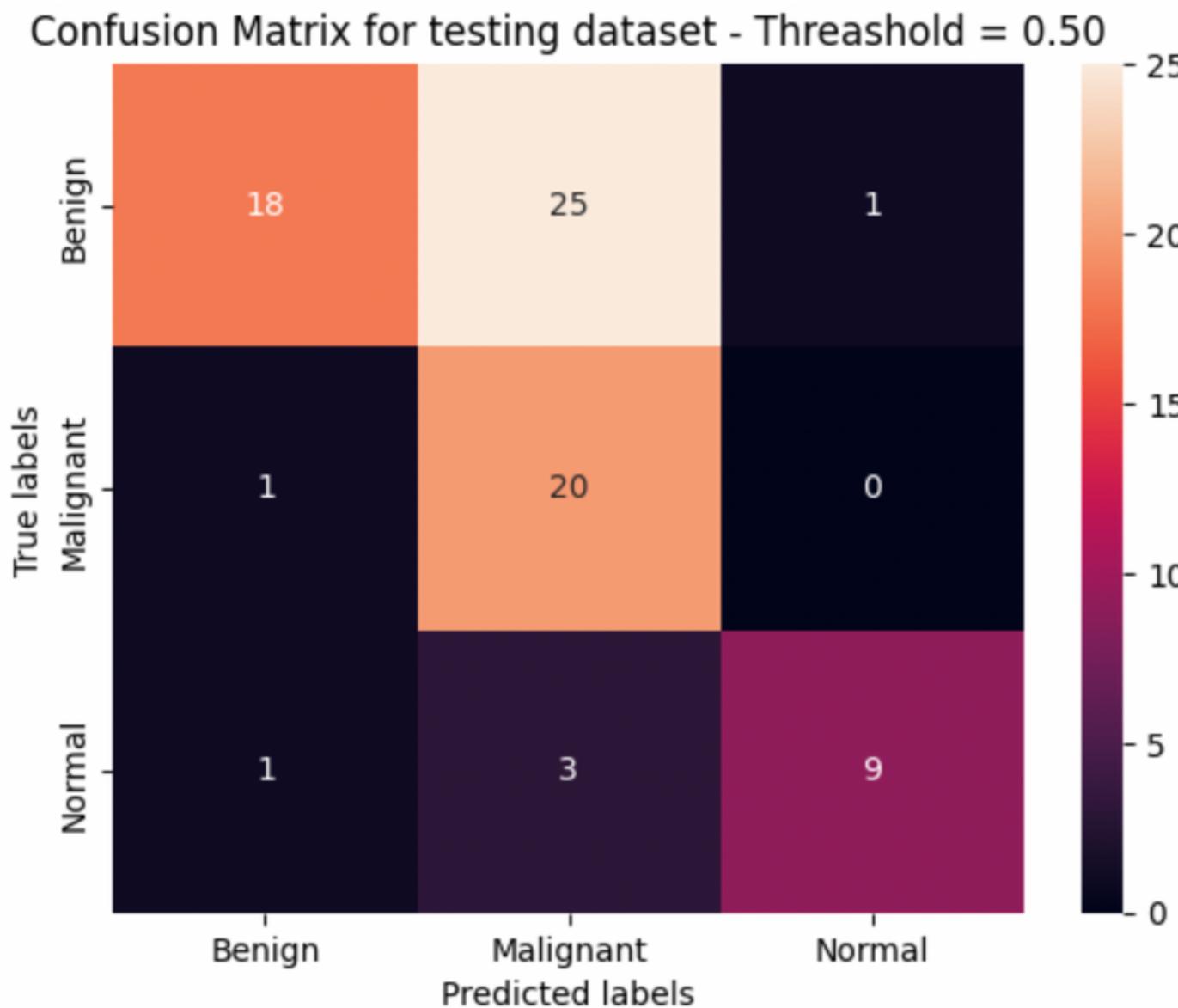


```
tensor([[-0.5234, -0.1519,  0.3111]]) = Malignant
tensor([[ 0.0104,  0.0114,  0.0009]]) = Normal
tensor([[-1.1133,  0.0298,  0.2126]]) = Benign
```

Résultats combinés - labels

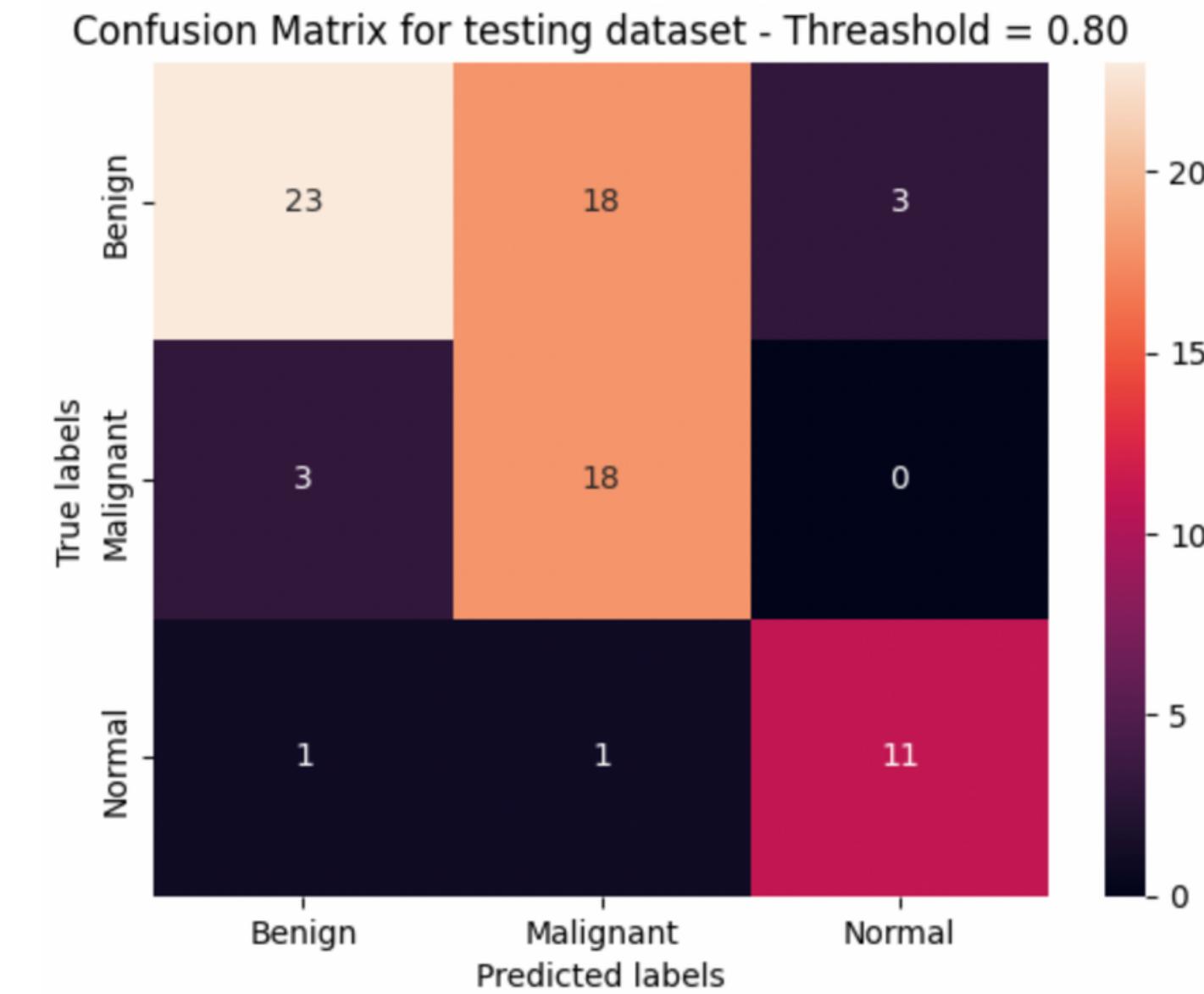
Test : changement du seuil d'arrondi des masques

Arrondi des masques avec un seuil à 0.5



Test Accuracy : 60,3%

Arrondi des masques avec un seuil à 0.8



Test Accuracy : 66,7%

Application d'une sigmoid sur l'output
et classement sur l'output

Résultats sur le set de test

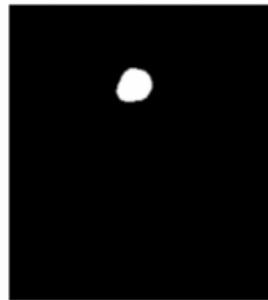
Ultrasound Image



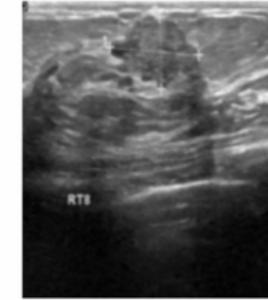
Target : benign



Prediction :benign



Ultrasound Image



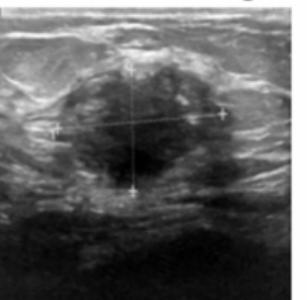
Target : benign



Prediction :malignant



Ultrasound Image



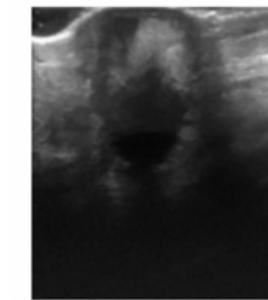
Target : malignant



Prediction :malignant



Ultrasound Image



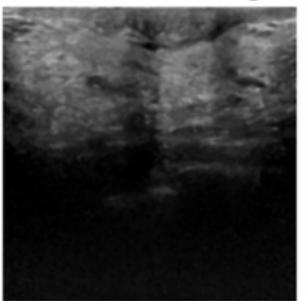
Target : malignant



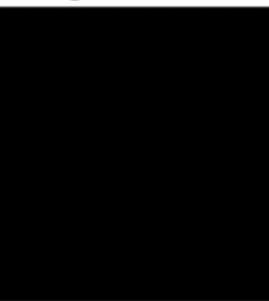
Prediction :malignant



Ultrasound Image



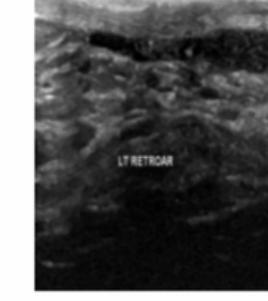
Target : normal



Prediction :normal



Ultrasound Image



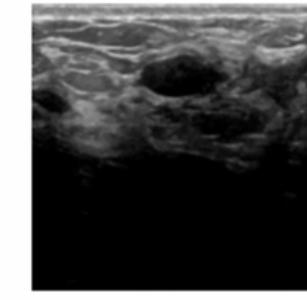
Target : benign



Prediction :normal



Ultrasound Image



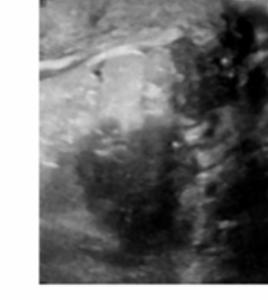
Target : benign



Prediction :benign



Ultrasound Image



Target : malignant



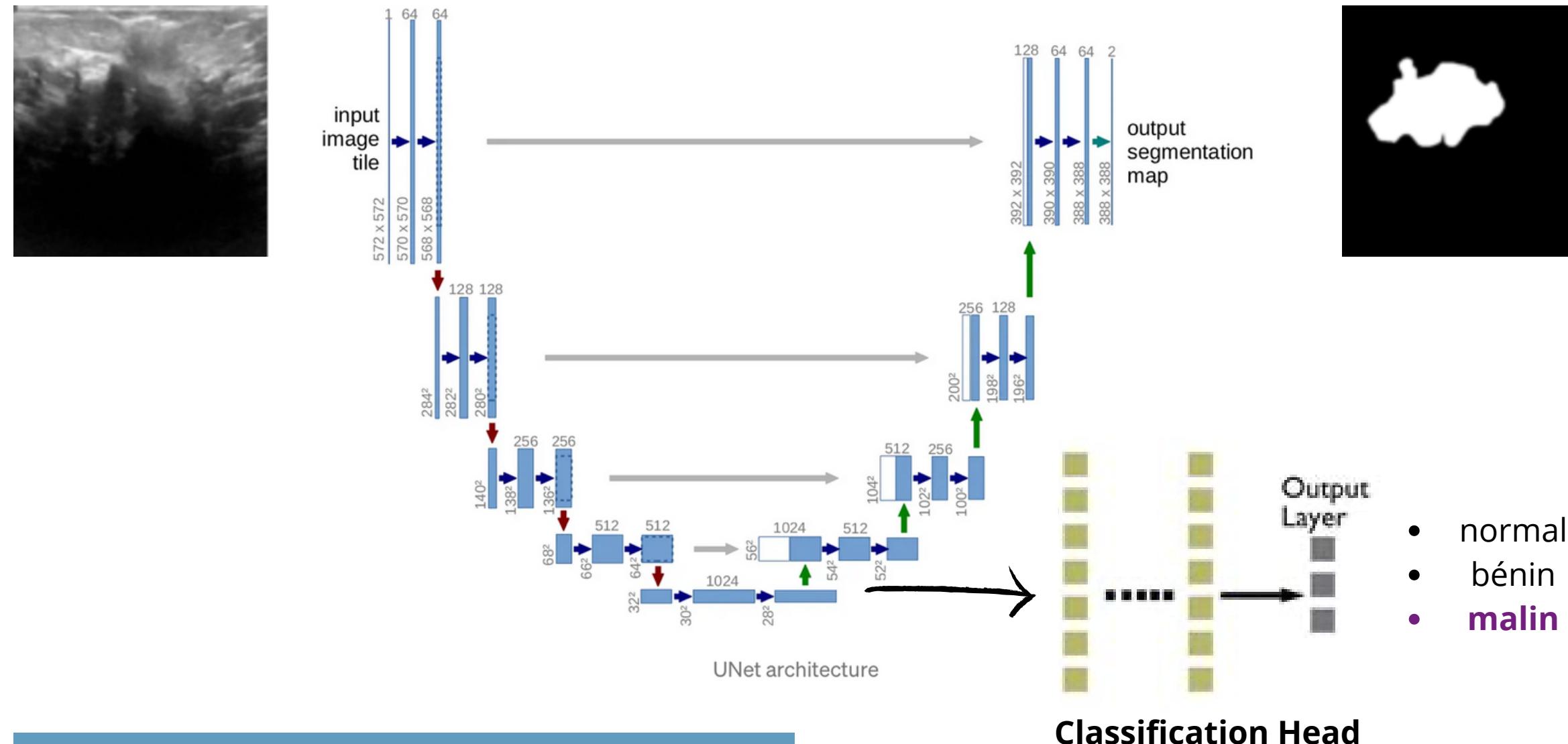
Prediction :malignant



02

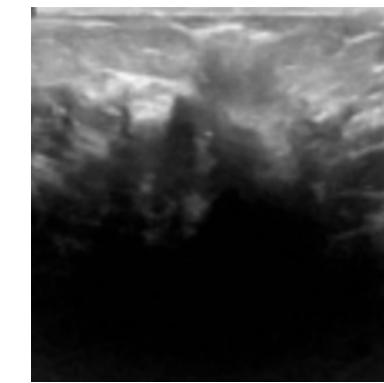
Modèle hybride

Architecture



Unet + classification head
encoder = resnet34
pre-trained sur ImageNet

Losses & Optimizers



- normal
- bénin
- **malin**

- Opt 1: Optimizer pour la segmentation
- Opt 2: Optimizer pour la classification

2 losses différentes pour chacune des tâches

- **MixedLoss** (mask_predit, mask_cible)
- **CrossEntropy** (label_predite, label_cible)

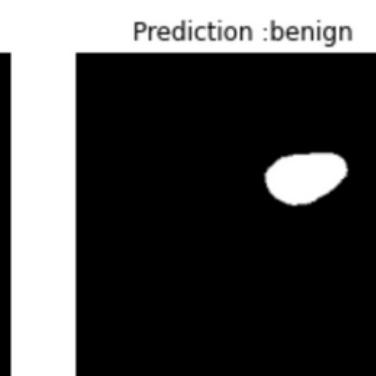
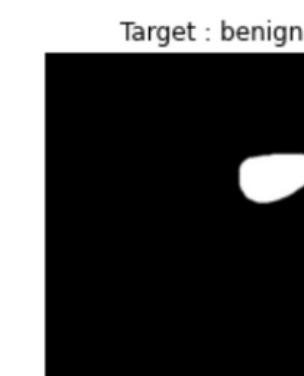
Accuracy et loss

Jeu de validation	Jeu de test
Accuracy : 88,7%	Accuracy : 80%
Cross Entropy moyenne : 0.070	Cross Entropy moyenne : 0.077
Mixed Loss moyenne : 0.158	Mixed Loss moyenne : 0.159

Résultats sur le set de test



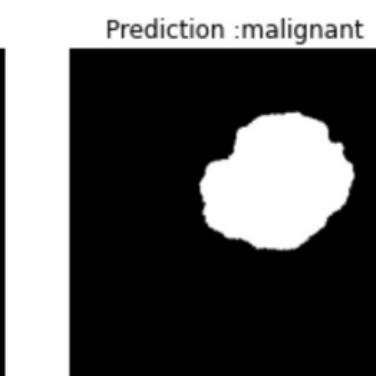
✓



✓



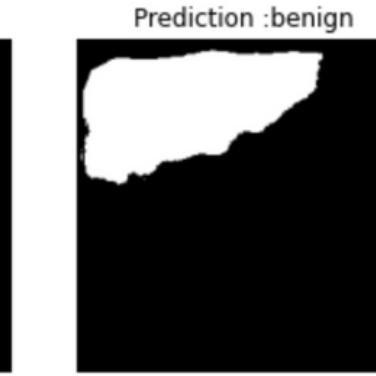
✓



✓



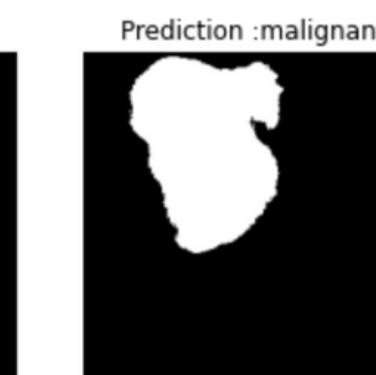
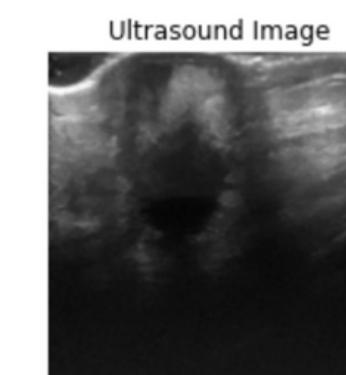
✓



✗

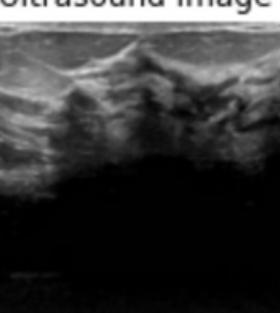
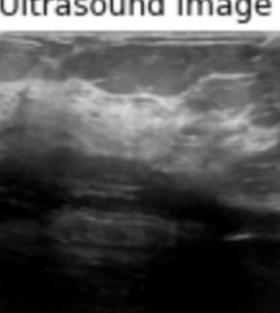
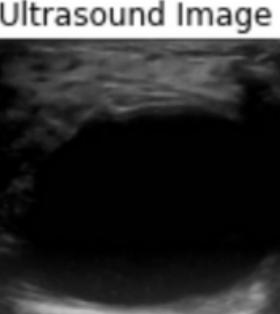
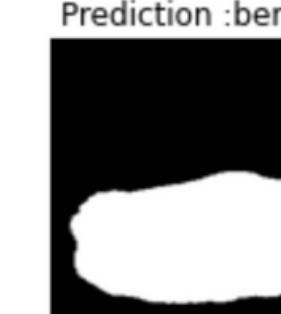
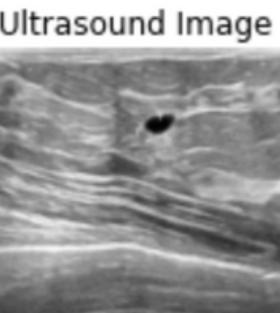
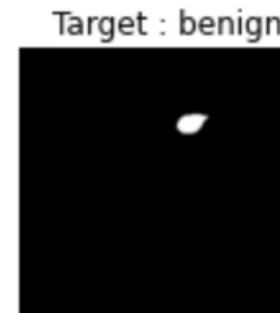
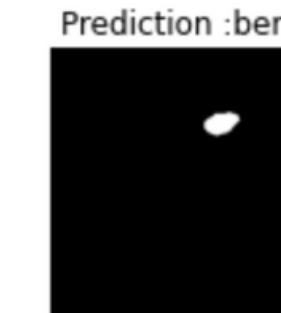


✗



✓

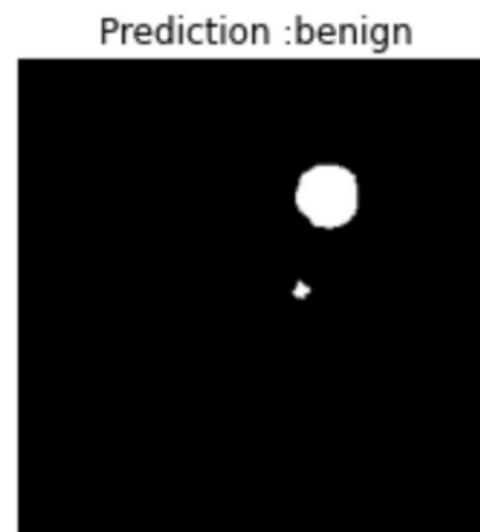
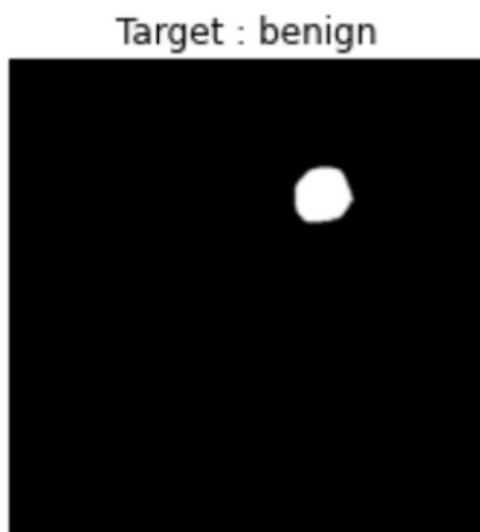
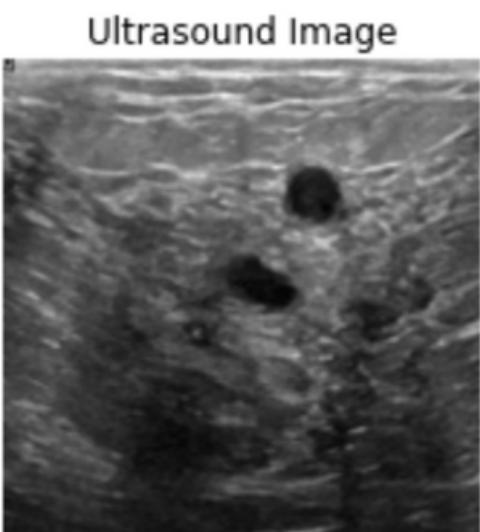
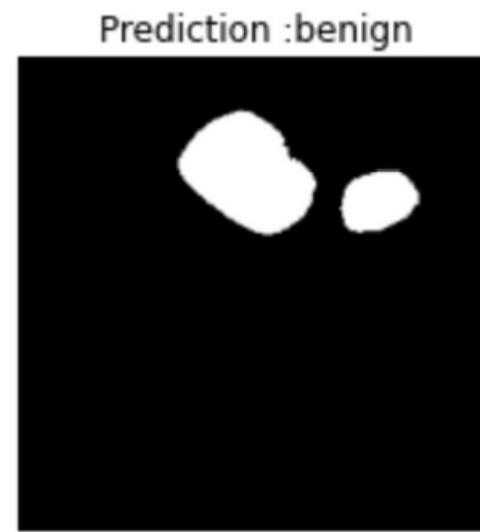
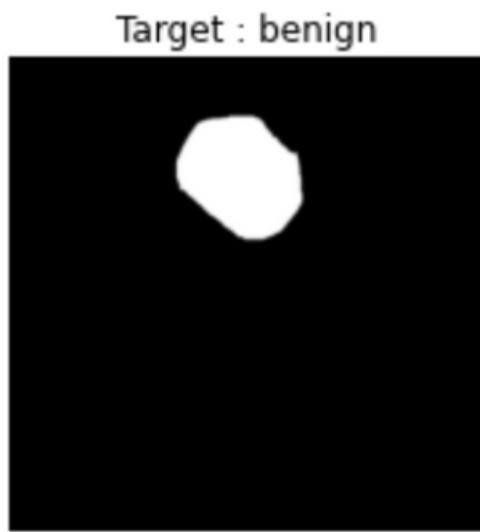
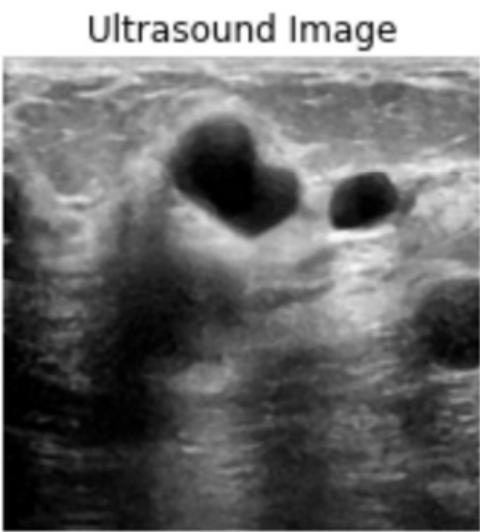
Résultats sur le set de test

Ultrasound Image 	Target : normal 	Prediction :normal 	✓	Ultrasound Image 	Target : malignant 	Prediction :benign 	✗
Ultrasound Image 	Target : normal 	Prediction :normal 	✓	Ultrasound Image 	Target : malignant 	Prediction :malignant 	✓
Ultrasound Image 	Target : benign 	Prediction :benign 	✓	Ultrasound Image 	Target : benign 	Prediction :benign 	✓
Ultrasound Image 	Target : benign 	Prediction :benign 	✓	Ultrasound Image 	Target : benign 	Prediction :benign 	✓

Résultats sur le set de test

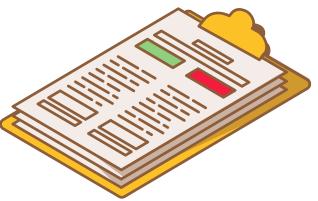
Ultrasound Image	Target : benign	Prediction : normal	X	Ultrasound Image	Target : malignant	Prediction : malignant	X
Ultrasound Image	Target : malignant	Prediction : malignant	✓	Ultrasound Image	Target : malignant	Prediction : benign	X
Ultrasound Image	Target : benign	Prediction : benign	✓	Ultrasound Image	Target : malignant	Prediction : benign	X
Ultrasound Image	Target : benign	Prediction : benign	✓	Ultrasound Image	Target : malignant	Prediction : benign	X

Cas intéressants



02

Conclusion



Conclusion

- Application des techniques de deep learning étudiées en cours
- Résultats satisfaisants pour un projet étudiant



Pistes d'amélioration

- Rééquilibrer le jeu de données
- Utiliser un autre set de test indépendant (autre pays, autre centre, autre échographe)
- Pénaliser différemment les erreurs : FN à éviter en priorité

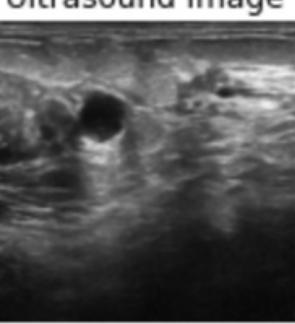
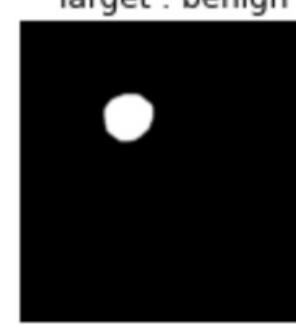
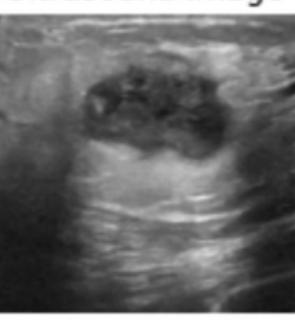
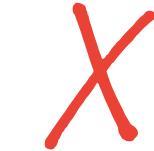
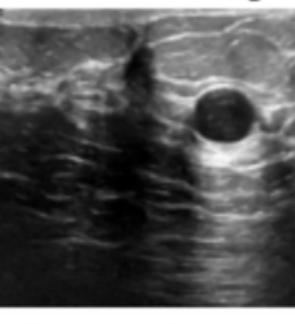
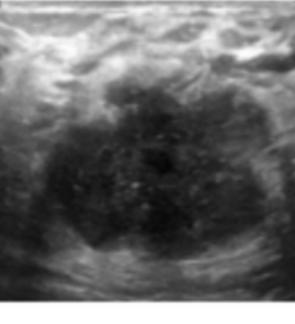
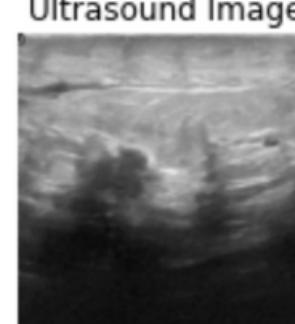
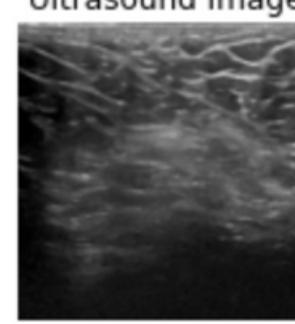
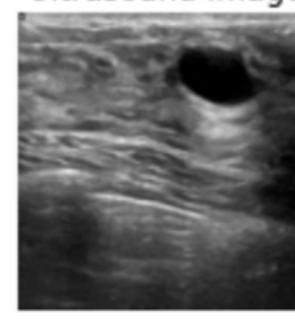
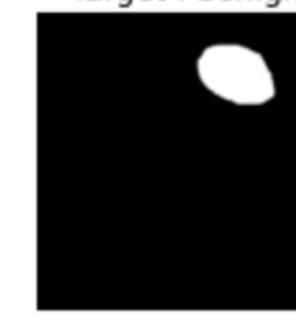
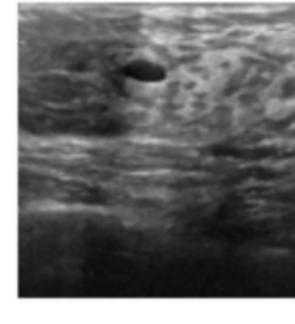


Idées d'applications

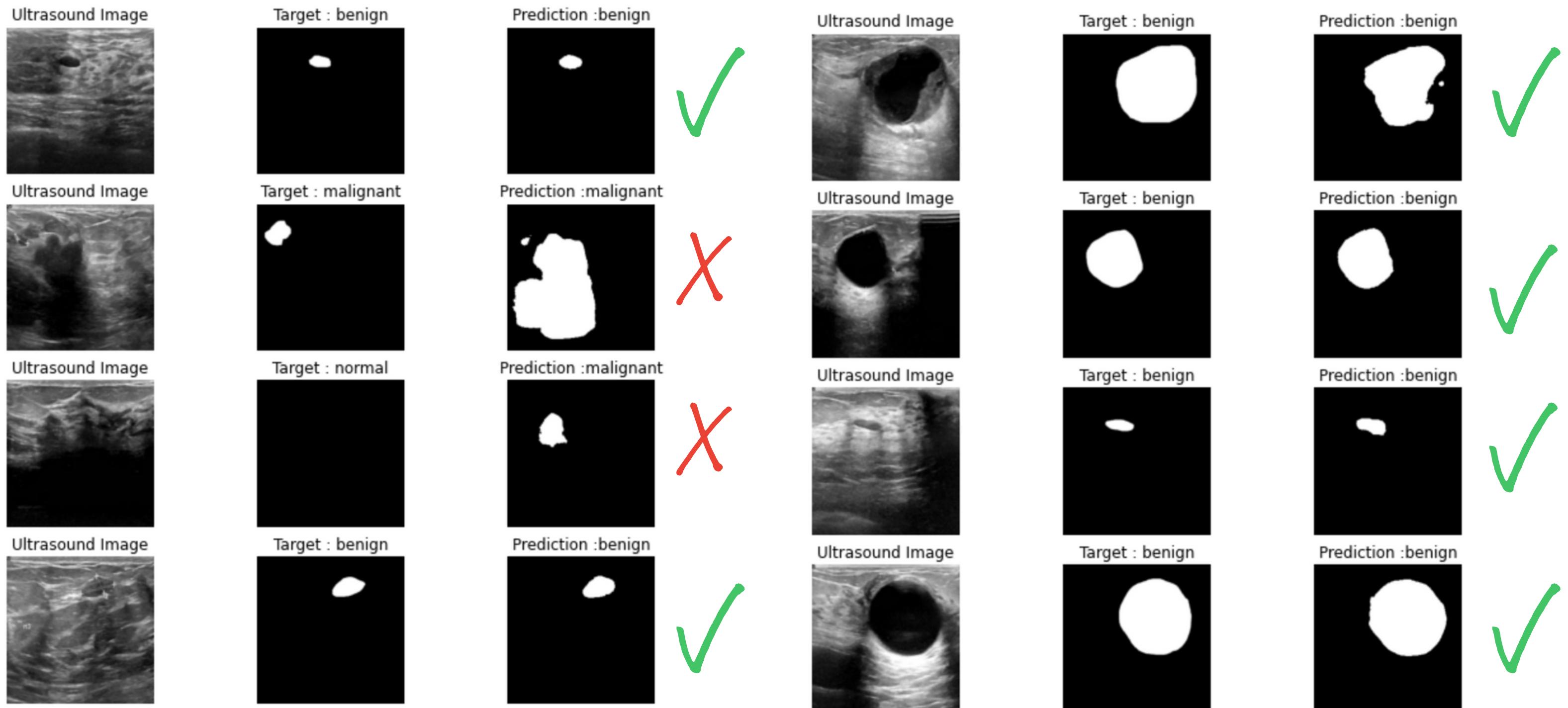
- Filet de sécurité + réassurance diagnostic
- Outils de formation des internes
- Échographe portable avec aide au diagnostic intégrée

Merci pour votre attention !
Des questions ?

Résultats sur le set de test

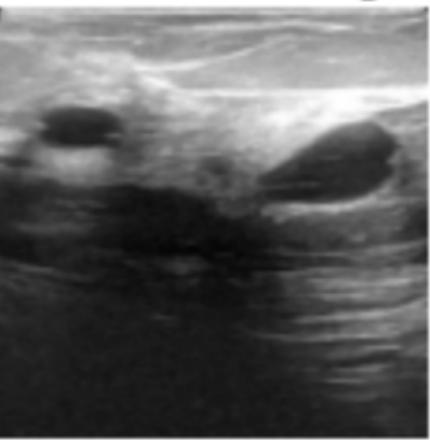
	Target : benign			
	Target : benign			
	Target : benign			
	Target : malignant			
	Target : malignant			
	Target : normal			
	Target : benign			
	Target : benign			

Résultats sur le set de test



Cas intéressants

Ultrasound Image



Target : benign



Prediction :benign



Ultrasound Image



Target : benign



Prediction :benign



Modele 2 steps : validation dataset

