

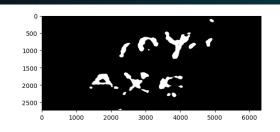
The goal is to detecting ink from 3D X-ray scans and reading the contents. Due to the heat of the volcano, the scrolls were carbonized, and are now impossible to open without breaking them. We will try to find the best model to get the maximum accuracy.

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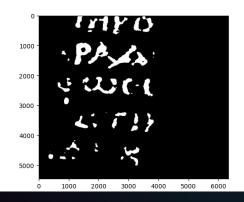
	Suits	LR	Loss	М	Optimizer	Score	Conclusions
	ResNet18	0.00001	BCEWithLo gitsLoss	0.6	Adam	0	Next page ——▶
	ResNet34	0.00001	BCEWithLo gitsLoss	0.6	Adam	0	Next page ───►
	ResNet50	0.0001	BCEWithLo gitsLoss	-	Adam	0	
	ResNet101	0.0001	BCEWithLo gitsLoss	0.9	Adam	Failed	
	ResNet152	0.0001	BCEWithLo gitsLoss	0.9	Adam	Failed	Overfitting after 3 epocs.
	ResNet101_32x _4d	0.0001	DiseLoss	0.9	Adam	0	
•	Efficientnet- b5	0.0001	BCEWithLo gitsLoss	0.9	Adam	0.28	

ResNet18

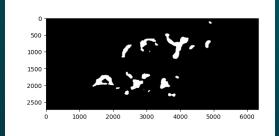


100%| | 13/13 [00:12<00:00, 1.04it/s] 100%| | 10805/10805 [45:28<00:00, 3.96it/s]

mask_count_min: 1.0

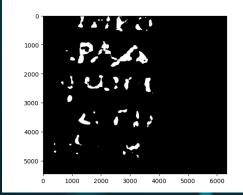


ResNet34



100%| | 13/13 [00:13<00:00, 1.01s/it] 100%| | 10805/10805 [48:50<00:00, 3.69it/s]

mask_count_min: 1.0



E	sults	LR	Loss	М	Optimizer	Score	Conclusions
	ResNet18	0.0001	BCEWithLo gitsLoss	-	Adam	0	
	ResNet34	0.0001	BCEWithLo gitsLoss	-	Adam	0	
	ResNet50	0.0001	BCEWithLo gitsLoss	-	Adam	Failed	
I	nception-V4	0.0001	BCEWithLo gitsLoss	-	Adam	Failed	Overfitting
	ResNet151	0.0001	BCEWithLo gitsLoss	-	Adam	0	-

Base Notebook

50%



30%

PAVEL HANCHAR . 1075TH IN THIS COMPETITION . POSTED A MONTH AGO

Color analysis: Ink = Ink + Noise. ROI

20%



2.5d segmentaion baseline [training]

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2.5d segmentaion baseline [inference]

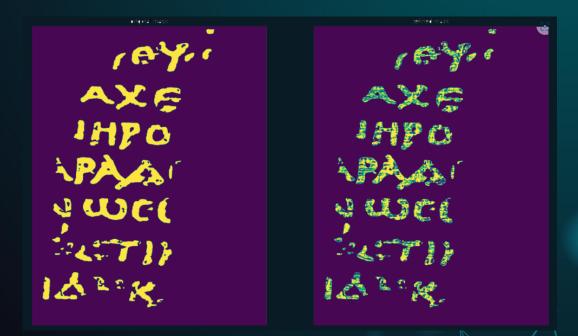
Notebook copied with edits from a private notebook · Updated 2mo ago Score: 0.41 · 29 comments · Vesuvius Challenge - Ink Detection +5

- Show the data
 Create Dataset Object
 Graph functions
- Gaussian distribution
 Take specific layers
 cutting boundary pixels
 Reset abnormal pixels

Fragment	Slices	Ink Peak
Fragment 1	21-34	65
Fragment 2	25-38	88
Fragment 3	20-33	77

Test and submission RLE

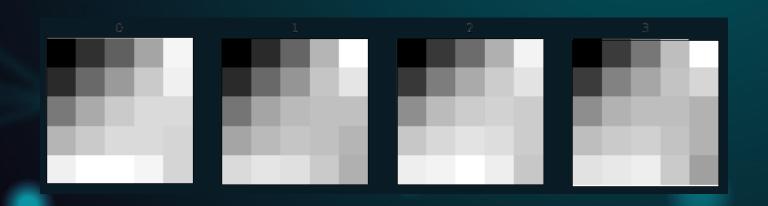
1 Reset abnormal pixels



2 Clipping Pixeles



3 augmentation that does the opposite of max polling



4 Focused on the inner part of the fragment.



Worked	Didn't work		
Creating the base notebook.	DiseLoss – Negative loss		
Clipping the fragment pixels to the correct ink peak.	The opposite of max polling – RAM Problem.		
Reset abnormal pixels	SKLearn Algorithm.		
ResNet18/34	ResNet101/152		
Local Submission (Visual text)	Submission (Score O/ Failed)		

What we learn?

3D Segmantations Volumes

Summ

Summary course

3 Impressions of models

4

Augmentataion