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PALMERAIE ROTANA RESORT
MARRAKESH / MOROCCO

PasteurDBC : Mednext adoption for fetal brain MRI segmentation

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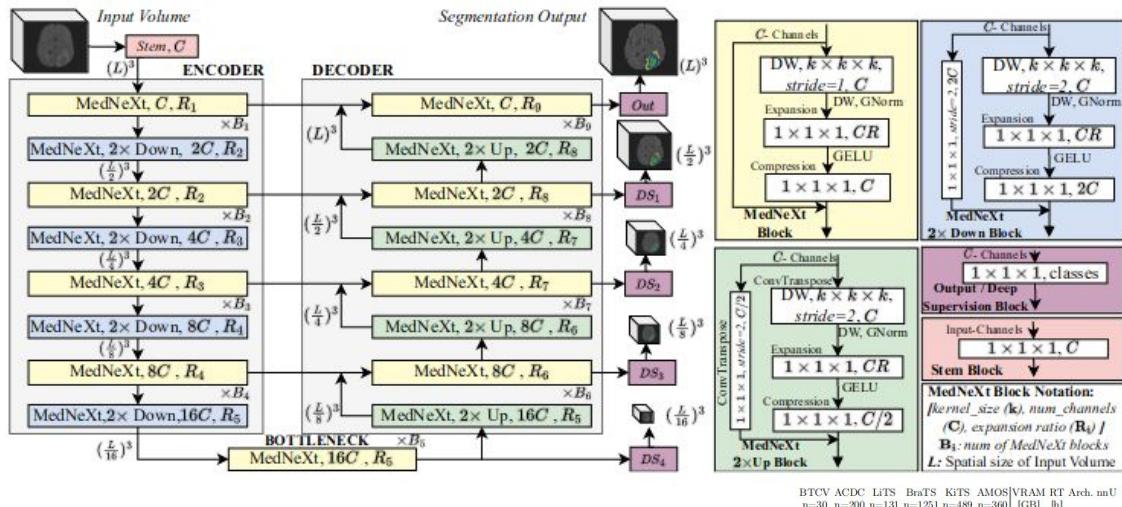
1-Decision and Bayesian Computation - Épiméthée
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Segmentation task : MedNeXt Model presentation

- Keeps the inductive bias of the convolutional layers
- Compares favorably with SOTA architectures, including transformers, when trained from scratch
- Is *de facto* integrated in the nnUnet segmentation pipeline



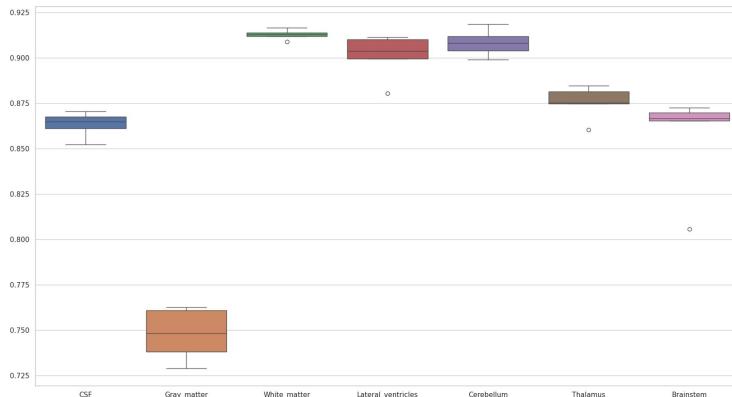
[1] F. Isensee, et al. "nnu-net revisited: A call for rigorous validation in 3d medical image segmentation." *arXiv preprint arXiv:2404.09556* (2024).

[2] S. Roy, et al. "Mednext: transformer-driven scaling of convnets for medical image segmentation." *International Conference on Medical Image Computing and Computer-Assisted Intervention*. Cham: Springer Nature Switzerland, 2023.

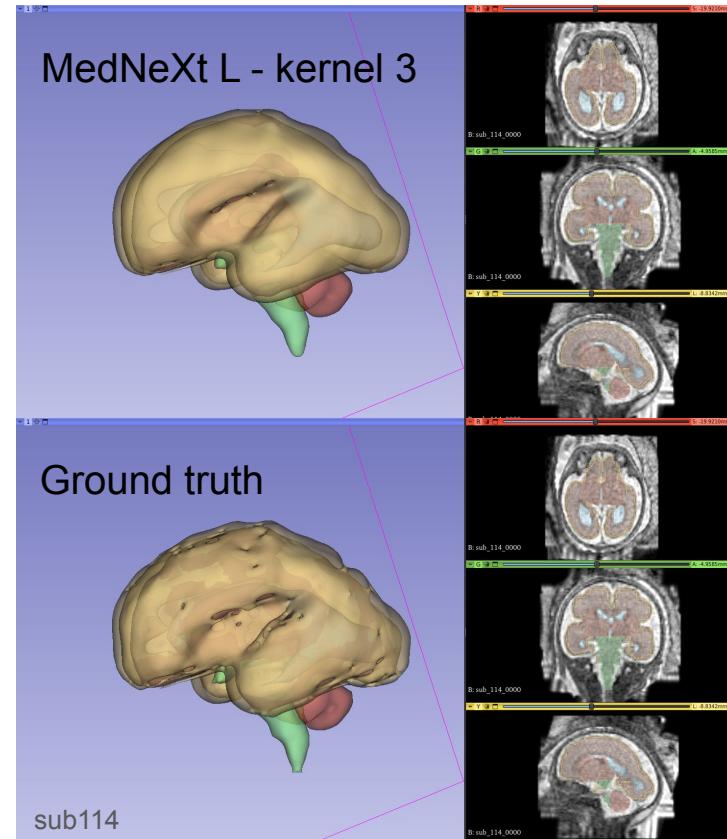
	BCTV	ACDC	LITS	Brats	KITS	AMOS	VRAM	RT	Arch.	mu
							[GB]	[h]		
nnU-Net (org.) [24]	83.08	91.54	80.09	91.24	86.04	88.64	7.70	9	CNN	Yes
nnU-Net ResEnc M	83.31	91.99	80.75	91.26	86.78	88.77	9.10	12	CNN	Yes
nnU-Net ResEnc L	83.35	91.99	81.60	91.13	88.17	89.41	22.70	35	CNN	Yes
nnU-Net ResEnc XL	83.28	91.48	81.19	91.18	88.67	89.68	36.60	66	CNN	Yes
MedNeXt L k3 [34]	84.70	92.65	82.14	91.35	88.25	89.62	17.30	68	CNN	Yes
MedNeXt L k5 [33]	85.04	92.62	82.34	91.50	87.74	89.73	18.00	233	CNN	Yes
STU-Net S [20]	82.92	91.04	78.50	90.55	84.93	88.08	5.20	10	CNN	Yes
STU-Net B [20]	83.05	91.30	79.19	90.85	86.33	88.46	8.80	15	CNN	Yes
STU-Net L [20]	83.36	91.31	80.31	91.26	85.84	89.34	26.50	51	CNN	Yes
SwinUNet [22]	78.89	91.29	76.50	90.68	81.27	83.81	13.10	15	TF	Yes
SwinUNETRV2 [25]	80.85	92.01	77.85	90.74	84.14	86.24	13.40	15	TF	Yes
nnFormer [21]	80.86	92.40	77.40	90.22	75.85	81.55	5.70	8	TF	Yes
CoTr [26]	81.95	90.56	79.10	90.73	84.59	88.02	8.20	15	TF	Yes
No-Mamba Base	83.69	91.89	80.57	91.26	85.96	89.04	12.0	24	CNN	No
No-Mamba Bot [20]	83.51	91.79	80.40	91.26	86.22	89.13	12.40	24	Mam	Yes
No-Mamba Enc [20]	82.41	91.22	80.27	90.91	83.34	88.38	24.90	47	Mam	Yes
A3DS SeqResNet [28]	80.69	90.69	70.28	90.79	81.11	87.27	20.00	22	CNN	No
A3DS DINTS [28]	78.18	82.97	60.05	87.75	65.28	82.35	29.20	16	CNN	No
A3DS SwinUNETR [29]	76.54	82.68	68.59	89.90	52.82	85.05	31.50	9	TF	No

Segmentation task : Results

- Predicts smoother structures than the expert annotations
- Compares on par with previous FeTA submissions
- Has been evaluated by radiologist experts on a private dataset : 45 reconstructed foetus MRI (1.5T)



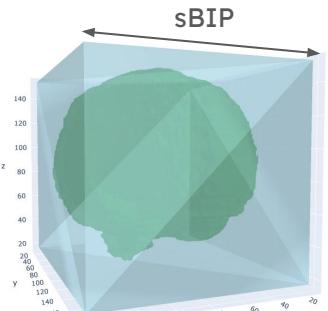
[3] K. Payette, et al. "Multi-Center Fetal Brain Tissue Annotation (FeTA) Challenge 2022 Results." *arXiv preprint arXiv:2402.09463* (2024).



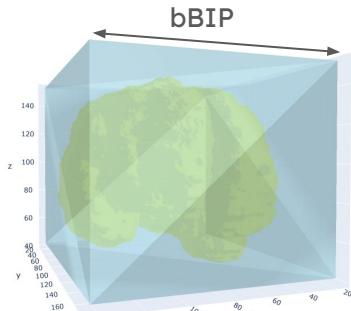
Biometry task : Rule based strategy

- Measures correlate well with segmentations bounding boxes
- Estimation of the rotation matrix would certainly improve the results

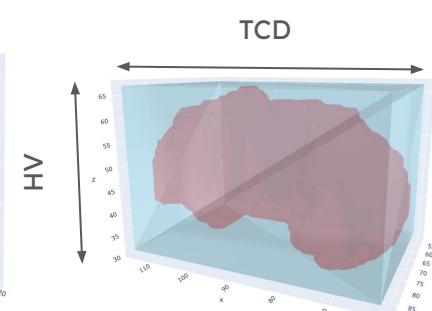
Measure	ME (%)	R2
sBIP	5.703417	0.865769
bBIP	7.969022	0.754252
HV	39.190029	-2.12880
TCD	4.48192	0.94305
LCC	19.55445	-0.73959



Extra axial CSF



Gray matter



Cerebellum

