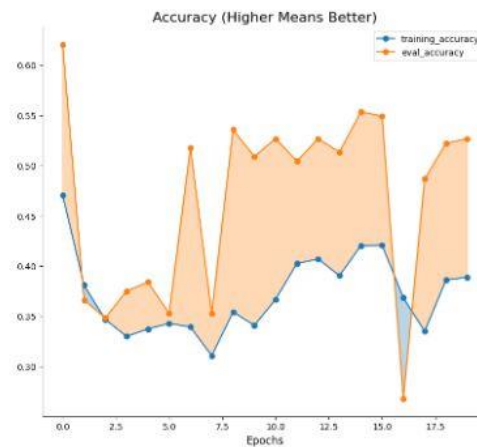
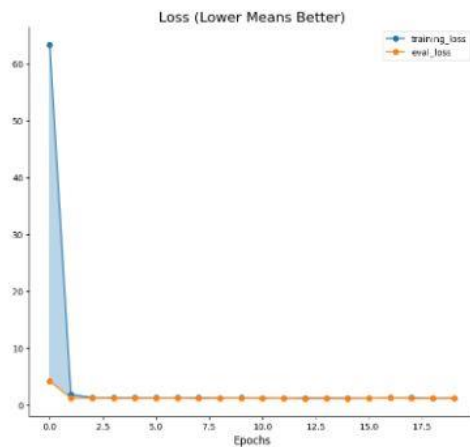


# Regularization Report

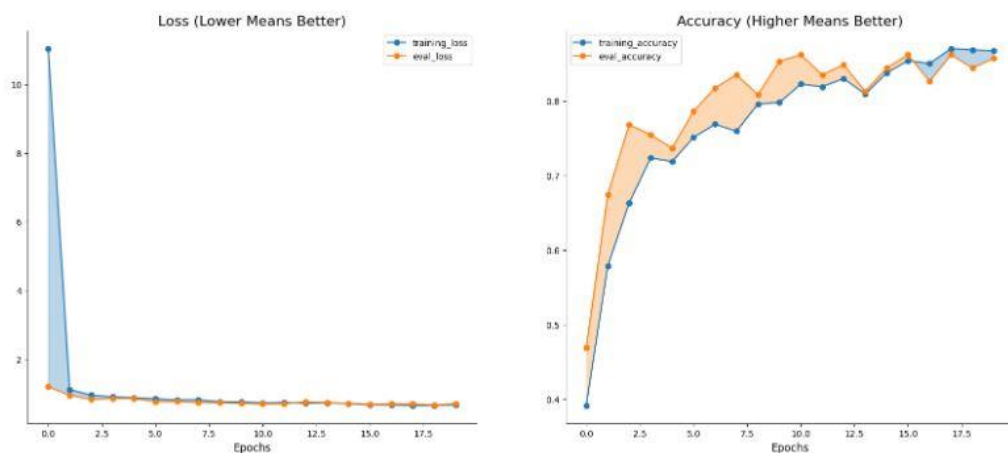
- Twice batch normalization that overfitted on pituitary class  
I think images are so dark or light .

	precision	recall	f1-score	support
glioma	0.00	0.00	0.00	64
meningioma	0.18	0.05	0.08	61
notumor	0.75	0.65	0.69	77
pituitary	0.34	0.98	0.50	61
accuracy			0.43	263
macro avg	0.31	0.42	0.32	263
weighted avg	0.34	0.43	0.34	263



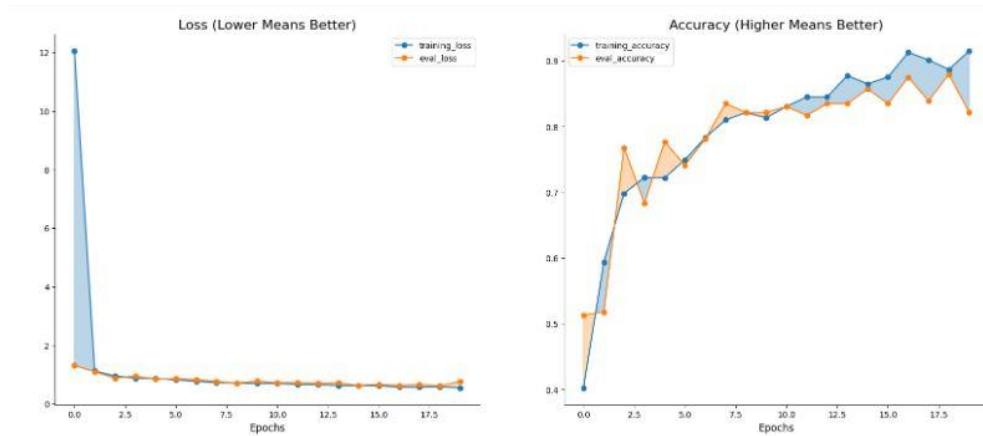
- Just one batch normalization layer for second convolution overfitting doesn't occurs but accuracy is less.

	precision	recall	f1-score	support
glioma	0.68	0.78	0.72	64
meningioma	0.51	0.43	0.46	61
notumor	0.86	0.87	0.86	77
pituitary	0.90	0.89	0.89	61
accuracy			0.75	263
macro avg	0.74	0.74	0.74	263
weighted avg	0.74	0.75	0.74	263



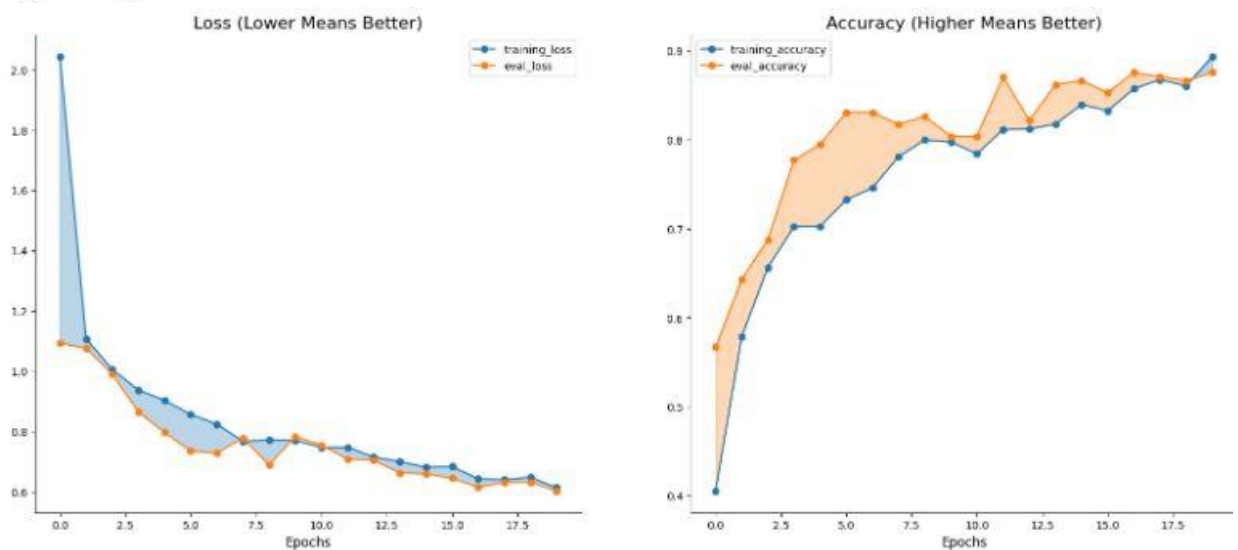
- Also by without drpout and batch normalization is not good enough .

	precision	recall	f1-score	support
glioma	0.90	0.56	0.69	64
meningioma	0.49	0.79	0.61	61
notumor	0.85	0.90	0.87	77
pituitary	0.89	0.66	0.75	61
accuracy			0.73	263
macro avg	0.78	0.73	0.73	263
weighted avg	0.79	0.73	0.74	263



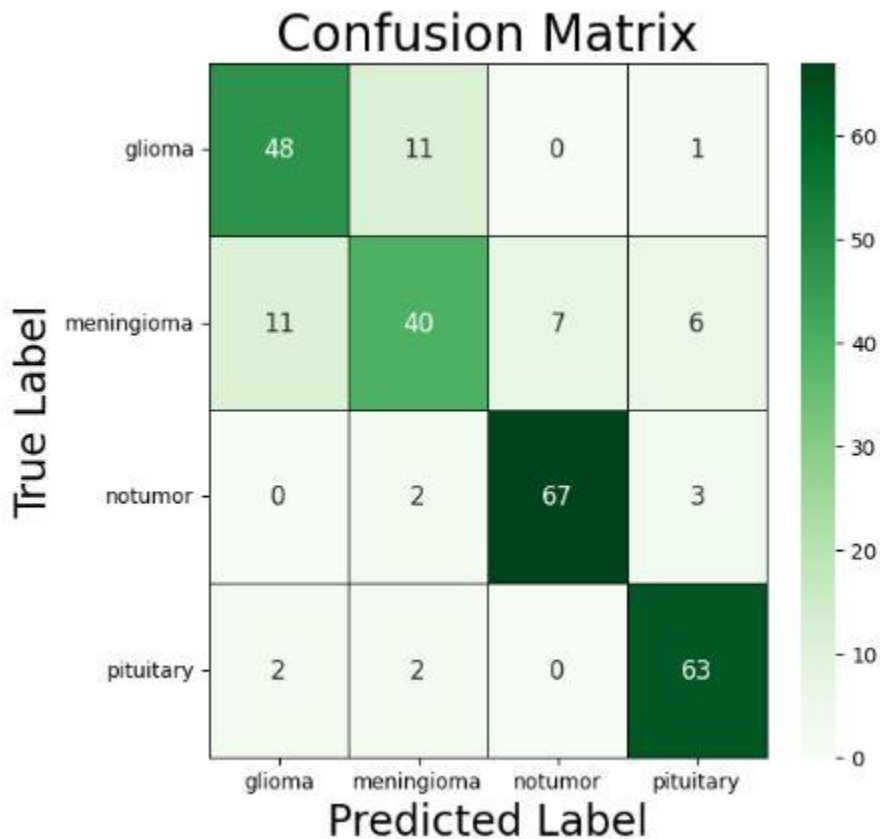
- By use one dropout with probability 0.5 I got the best result

	precision	recall	f1-score	support
glioma	0.82	0.70	0.76	64
meningioma	0.72	0.75	0.74	61
notumor	0.96	0.87	0.91	77
pituitary	0.80	0.97	0.87	61
accuracy			0.83	263
macro avg	0.82	0.82	0.82	263
weighted avg	0.83	0.83	0.82	263



# Final analysis

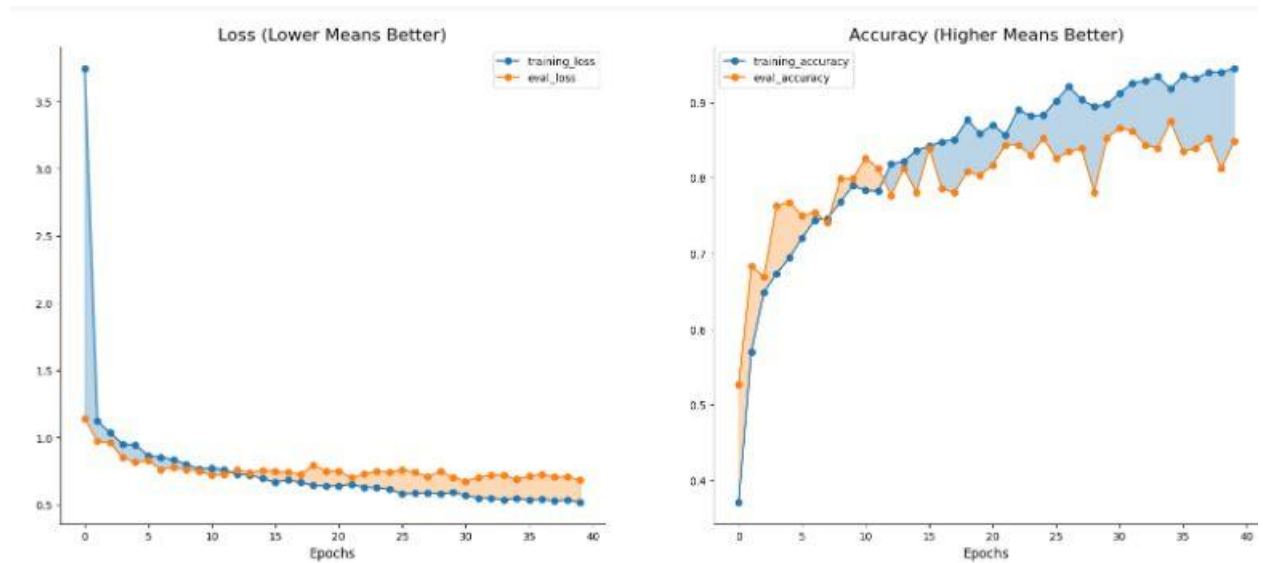
	precision	recall	f1-score	support
glioma	0.79	0.80	0.79	60
meningioma	0.73	0.62	0.67	64
notumor	0.91	0.93	0.92	72
pituitary	0.86	0.94	0.90	67
accuracy			0.83	263
macro avg	0.82	0.82	0.82	263
weighted avg	0.82	0.83	0.83	263



As we see final accuracy by f1-score is 0.83 .

Reason of lower accuracy is meningioma class that confused by glioma class because similarity between some samples.

Notumor is so good because data of this part is so visible and detectable.



Loss curves shows validation is a little worse than train because it use for check performance , not for learning.

Curves for train and validation are near and loss for train and validation is low s we can say we have no bias and variance so we don't have overfitting and underfitting.