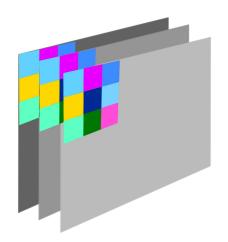
Involution: Results Presentation

Group 4

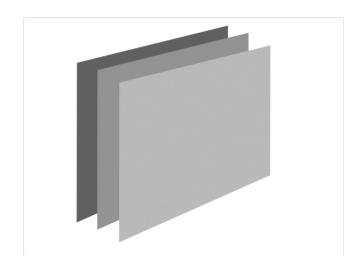
- 1.Harsh Chalikwar 2021A3PS2878G
- 2.Aditya Kurande 2021AAPS2485G
- 3.Atharv Mane 2020A7TS0153G
- 4. Simran Srivastava 2021AAPS2931G
- 5. Milind Kumar Prasad 2020A7PS0130G

Two key properties of Involution

Channel-Specific Agnostic



Spatial-Agnostic Specific



Code

https://github.com/thatblueboy/involution

```
class ReDSNet(nn.Module):
 arch settings = {
   26: (Bottleneck, (1, 2, 4, 1)),
   38: (Bottleneck, (2, 3, 5, 2)),
   50: (Bottleneck, (3, 4, 6, 3)),
   101: (Bottleneck, (3, 4, 23, 3)),
   152: (Bottleneck, (3, 8, 36, 3))
 def init (self,
               depth,
             in channels=3,
             stem channels=64,
             base channels=64,
             expansion = None,
             num stages=4,
             strides=(1, 2, 2, 2),
             out indices=(3, ),
             frozen stages=-1,
             avg down=False,
             zero init residual=True,
             is rednet = False,
             dropout = 0.1
```

```
ResNet-26 —> RedNet-26
ResNet-38 —> RedNet-38
ResNet-50 —> RedNet-50
ResNet-101 —> RedNet-101
ResNet-152 —> RedNet-152
```

```
from models.backbones import ReDSNet
rednet26 = ReDSNet(depth = 26, is_rednet=True)
```

Classification Head

```
class Classifier(nn.Module):
    def init (self, input feats, output feats):
        super(Classifier, self). init ()
       mid feats = (input feats+output feats)//2
        self.add module("classifier", nn.Sequential(
                nn.Linear(in features=input feats, out features=mid feats),
                nn.Linear(in features=mid feats, out features=output feats),
    def forward(self, x):
        logits = self.classifier(x)
        return logits
```

Pytorch Lightning



Name (126 visualized)	State	Notes	Use	Tag:	Crea ▼	Runtin	Sweep	dednet	dropou	is_redr	lr_sch€	lr_sch€	lr_sch€	num_c	optimi:	optimi:
model= <class 'models.rednet_clas<="" td=""><td>⊙ Failed</td><td>Add notes</td><td>f20200:</td><td></td><td>1d ago</td><td>4h 57m 1s</td><td></td><td>8-8</td><td>0.1</td><td>=</td><td>CosineAnr</td><td>130</td><td>0.00001</td><td>257</td><td>SGD</td><td>0.8</td></class>	⊙ Failed	Add notes	f20200:		1d ago	4h 57m 1s		8-8	0.1	=	CosineAnr	130	0.00001	257	SGD	0.8
model= <class 'models.resnet_clas<="" td=""><td>⊙ Finished</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>3h 45m 28</td><td>121</td><td>-</td><td>0.1</td><td>2</td><td>CosineAnr</td><td>130</td><td>0</td><td>257</td><td>SGD</td><td>0.8</td></class>	⊙ Finished	Add notes	f20200:		2d ago	3h 45m 28	121	-	0.1	2	CosineAnr	130	0	257	SGD	0.8
imagenet_tiny-type=50-train-bs=9	⊙ Finished	Add notes	thatblu		2d ago	53m 29s	-	9-	5 3	true	-	130	0	200	SGD	0.012
model= <class 'models.resnet_clas<="" td=""><td>⊙ Finished</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>2h 23m 13</td><td>(2)</td><td>-</td><td>0.1</td><td>2</td><td>CosineAnr</td><td>270</td><td>0</td><td>257</td><td>SGD</td><td>0.8</td></class>	⊙ Finished	Add notes	f20200:		2d ago	2h 23m 13	(2)	-	0.1	2	CosineAnr	270	0	257	SGD	0.8
model= <class 'models.resnet_clas<="" td=""><td>⊙ Finished</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>21m 32s</td><td>(E)</td><td>10-1</td><td>0.1</td><td>5</td><td>CosineAnr</td><td>270</td><td>0</td><td>257</td><td>SGD</td><td>0.0001</td></class>	⊙ Finished	Add notes	f20200:		2d ago	21m 32s	(E)	10-1	0.1	5	CosineAnr	270	0	257	SGD	0.0001
Imagenet_tiny-type=50-dropout-c	⊙ Finished	Add notes	thatblu		2d ago	53m 24s			9	true	(2)	130	0	200	SGD	0.012
● imagenet_tiny-type=50-bs=96- <cla< td=""><td>⊙ Finished</td><td>Add notes</td><td>thatblu</td><td></td><td>2d ago</td><td>9m 21s</td><td>-</td><td>10-1</td><td>=3</td><td>true</td><td>-</td><td>130</td><td>0</td><td>200</td><td>SGD</td><td>0.012</td></cla<>	⊙ Finished	Add notes	thatblu		2d ago	9m 21s	-	10-1	=3	true	-	130	0	200	SGD	0.012
model= <class 'models.resnet_clas<="" td=""><td>⊙ Finished</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>3h 51m 22</td><td>-</td><td>12</td><td>0.1</td><td>2</td><td>CosineAnr</td><td>130</td><td>0</td><td>257</td><td>SGD</td><td>0.8</td></class>	⊙ Finished	Add notes	f20200:		2d ago	3h 51m 22	-	12	0.1	2	CosineAnr	130	0	257	SGD	0.8
imagenet_tiny-type=50-dropout-b	⊙ Finished	Add notes	thatblu		2d ago	1h 33m 59	-	-	2 3	true	-	130	0	200	SGD	0.012
model= <class 'models.resnet_clas<="" td=""><td>⊙ Finished</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>12s</td><td>•</td><td>101</td><td>9</td><td>1</td><td>(U)</td><td>-</td><td>2</td><td>2</td><td>-</td><td>-</td></class>	⊙ Finished	Add notes	f20200:		2d ago	12s	•	101	9	1	(U)	-	2	2	-	-
model= <class 'models.rednet_clas<="" td=""><td>⊙ Failed</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>21m 10s</td><td></td><td>-</td><td>0.3</td><td>5</td><td>CosineAnr</td><td>130</td><td>0</td><td>257</td><td>SGD</td><td>0.8</td></class>	⊙ Failed	Add notes	f20200:		2d ago	21m 10s		-	0.3	5	CosineAnr	130	0	257	SGD	0.8
model= <class 'models.rednet_clas<="" td=""><td>⊙ Failed</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>42m 50s</td><td>·</td><td>121</td><td>0.1</td><td></td><td>CosineAnr</td><td>130</td><td>0</td><td>257</td><td>SGD</td><td>0.8</td></class>	⊙ Failed	Add notes	f20200:		2d ago	42m 50s	·	121	0.1		CosineAnr	130	0	257	SGD	0.8
model= <class 'models.rednet_clas<="" td=""><td>⊙ Finished</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>22s</td><td></td><td>-</td><td>0.1</td><td>5</td><td>CosineAnr</td><td>130</td><td>0</td><td>257</td><td>1-100 ▼ 0</td><td>of 126 〈 〉</td></class>	⊙ Finished	Add notes	f20200:		2d ago	22s		-	0.1	5	CosineAnr	130	0	257	1-100 ▼ 0	of 126 〈 〉
● Caltech256-type=121-bs=4- <class '<="" td=""><td>⊙ Finished</td><td>Add notes</td><td>f20200:</td><td></td><td>2d ago</td><td>36s</td><td>121</td><td>121</td><td>0.1</td><td>u u</td><td>CosineAnr</td><td>130</td><td>0</td><td>257</td><td>SGD</td><td>0.8</td></class>	⊙ Finished	Add notes	f20200:		2d ago	36s	121	121	0.1	u u	CosineAnr	130	0	257	SGD	0.8

Caltech 256



- Caltech-256 is an object recognition dataset containing 30,607 real-world images, of different sizes, spanning 257 classes (256 object classes and an additional clutter class).
- Each class is represented by at least 80 images.

Data Augmentation

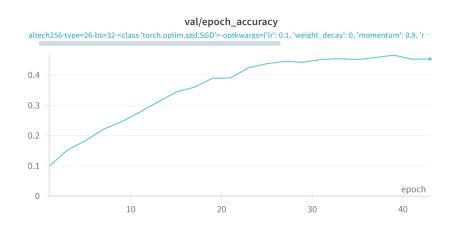
- Random horizontal flip with a probability of 0.5
- Random vertical flip with a probability of 0.5
- Random rotation by up to 45 degrees
- Random affine transformation by up to 45 degrees
- Normalize the image using mean=[0.485, 0.456, 0.406] and std=[0.229, 0.224, 0.225]

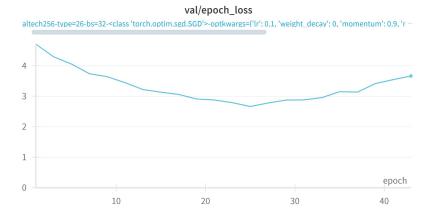
weight_decay: 0.0 grad_clip: None



Validation loss increasing while Training loss decreases -

Classic Overfitting!

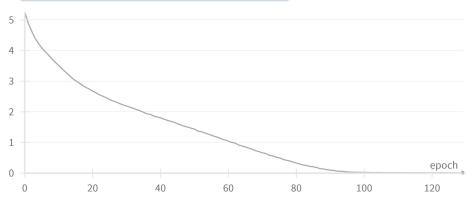




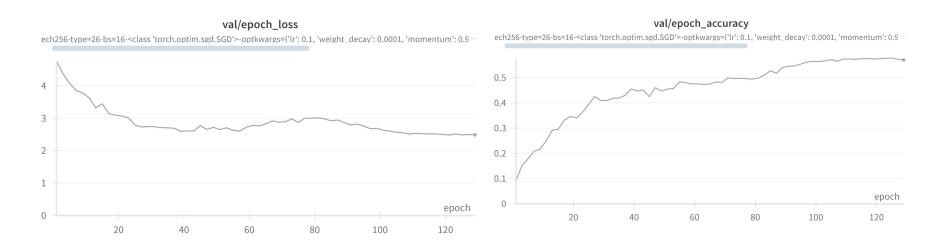
weight_decay: 1e-4 grad_clip: None epochs: 130

train/epoch_loss

ech256-type=26-bs=16-<class 'torch.optim.sgd.SGD'>-optkwargs={'lr': 0.1, 'weight_decay': 0.0001, 'momentum': 0.9 -



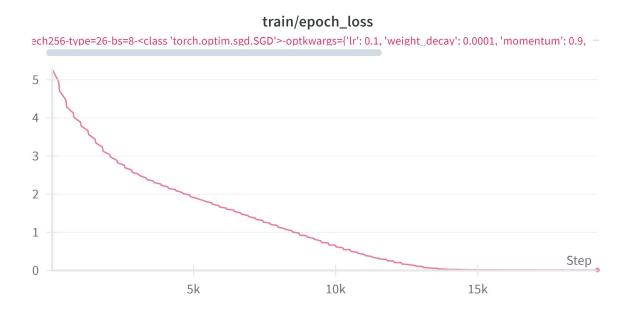
Validation loss Decreases and Converges while Training loss converges - Overfitting Solved!

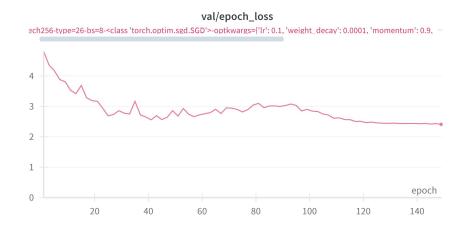


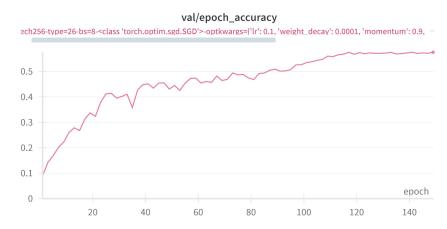
Final Accuracy for the Run: 57.03%

Benchmarking ResNet - 26 on Caltech: Longer Runs

weight_decay: 1e-4 grad_clip: None epochs: 150





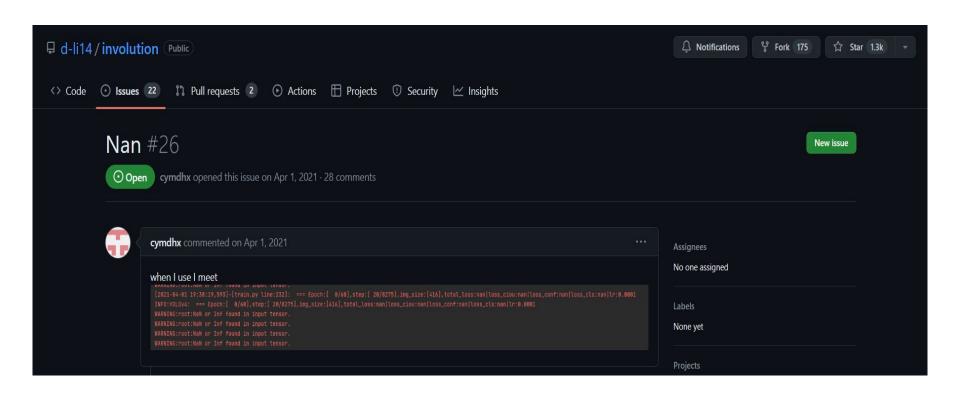


Final Accuracy for the Run: 57.59%

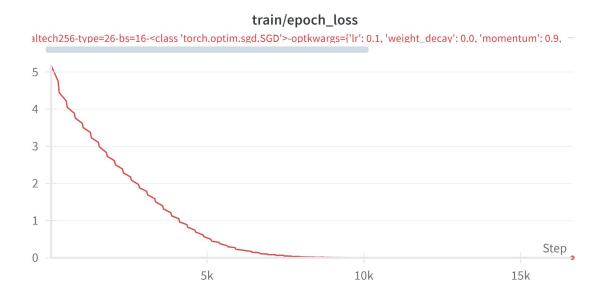
RedNet-50 on Caltech 256 - Another experiment to Validate NaNs - Some Internal Tensors Explode



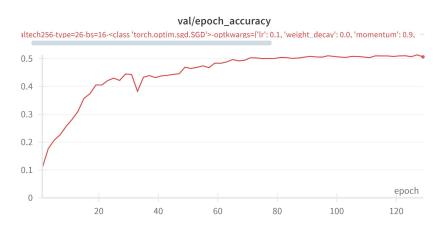
grad_clip: 0

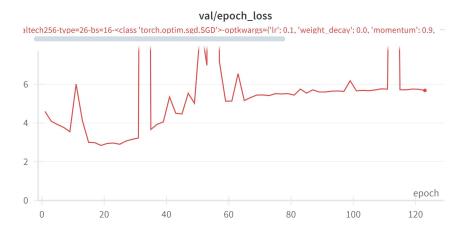


weight_decay: 0.0 grad_clip: None epochs: 130



Note: Instability can be observed in validation loss. We think this is again because of **Activations exploding**. The rough shape of the curve is same as that for Resnets but with abnormalities coming in between. Also note that validation loss has increasing trend. i.e. overfitting.



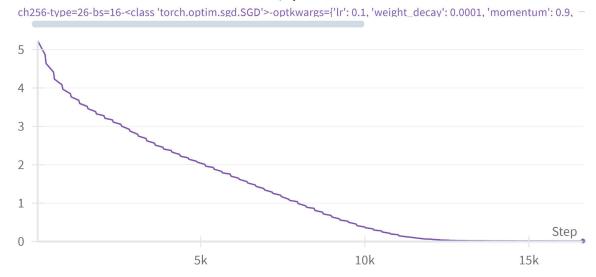


Final Accuracy for the Run: 50.6%

Optimiser: SGD **Ir_scheduler:** CosineAnnealingLR **Ir =** 0.1

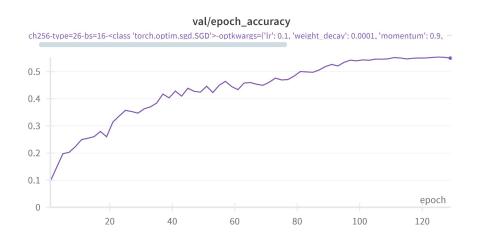
weight_decay: 1e-4 grad_clip: None epochs: 130 dropout: 10%

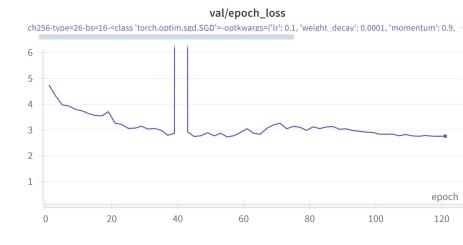
train/epoch_loss



Note: Irregularities in validation loss have significantly decreased. Final Accuracy Increases.

Validation loss reduces then stabilizes. Another point to note is that activation explosion does not happen for first few validation epochs. I.E. it is learnt.

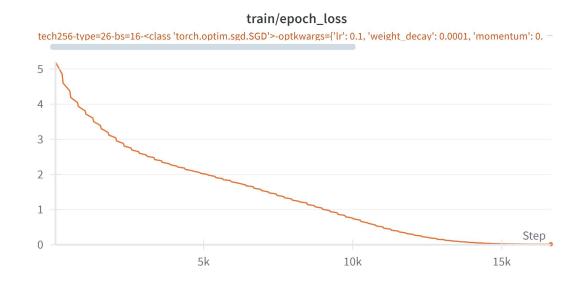




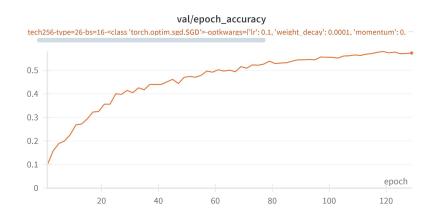
Final Accuracy for the Run: 55.00%

weight_decay: 1e-4 grad_clip: None epochs: 130 dropout: 10%

gradient clipping: 1



Note: Much stabler validation loss curve

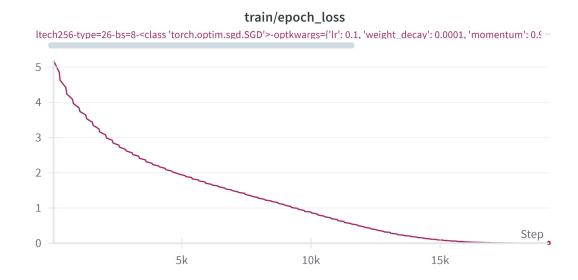




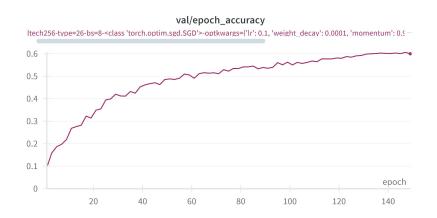
Final Accuracy for the Run: 57.3%

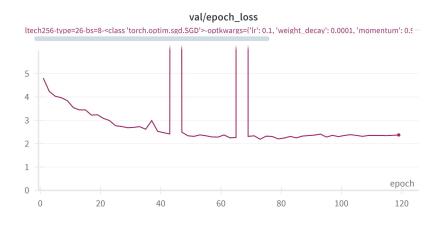
weight_decay: 1e-4 grad_clip: None epochs: 150 dropout: 10%

gradient clipping: 1



Note: Irregularities reappear for longer runs even with gradient clipping. This could be because of larger learning rate since the cosine annealing reduces the learning rate slower for longer runs. Nevertheless, accuracy increases.

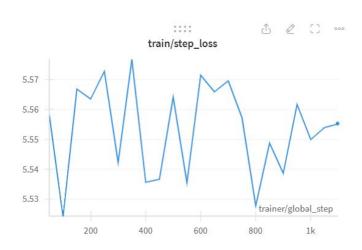


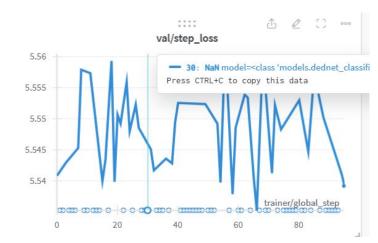


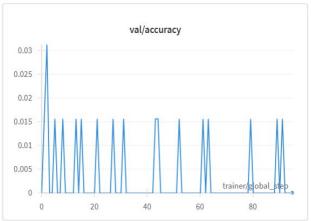
Final Accuracy for the Run: 60.05%

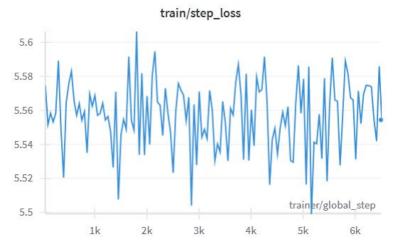
Model	No. Of Parameters	Accuracy
ResNet-26	11.7 M	57.59
RedNet-26	8 M	60.05

DEDNet: DenseNets with Involution









Note that the irregularities occur during "Sanity Check" runs i,e, default "kaiming initialization". This suggests that involution in and of itself might be a method with high instability. Which gets exaggerated in densenets



Thank You