

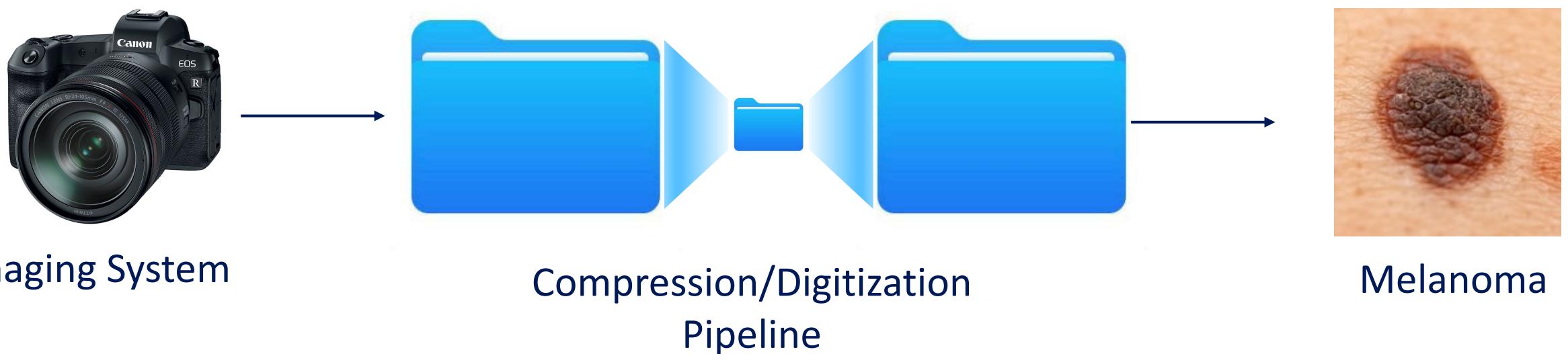
Insights From HAM10000 Classification Using Learned Compression

Yesh Doctor

BME548

Duke

Motivation



Hypothesis

- Learned compression can educate both imaging system design and image classification architectures

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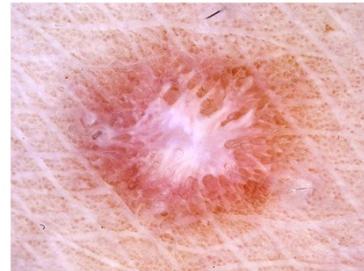
Dataset

Human Against Machine (HAM)-10000

Nevus



Dermatofibroma



Melanoma



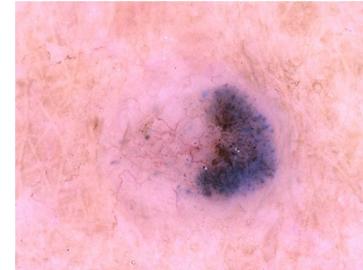
Pigmented
Bowen's



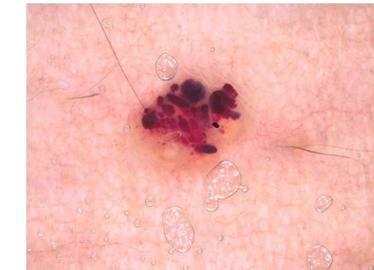
Pigmented Benign
Keratoses



Basal Cell
Carcinoma



Vascular



10015 Dermatological Images

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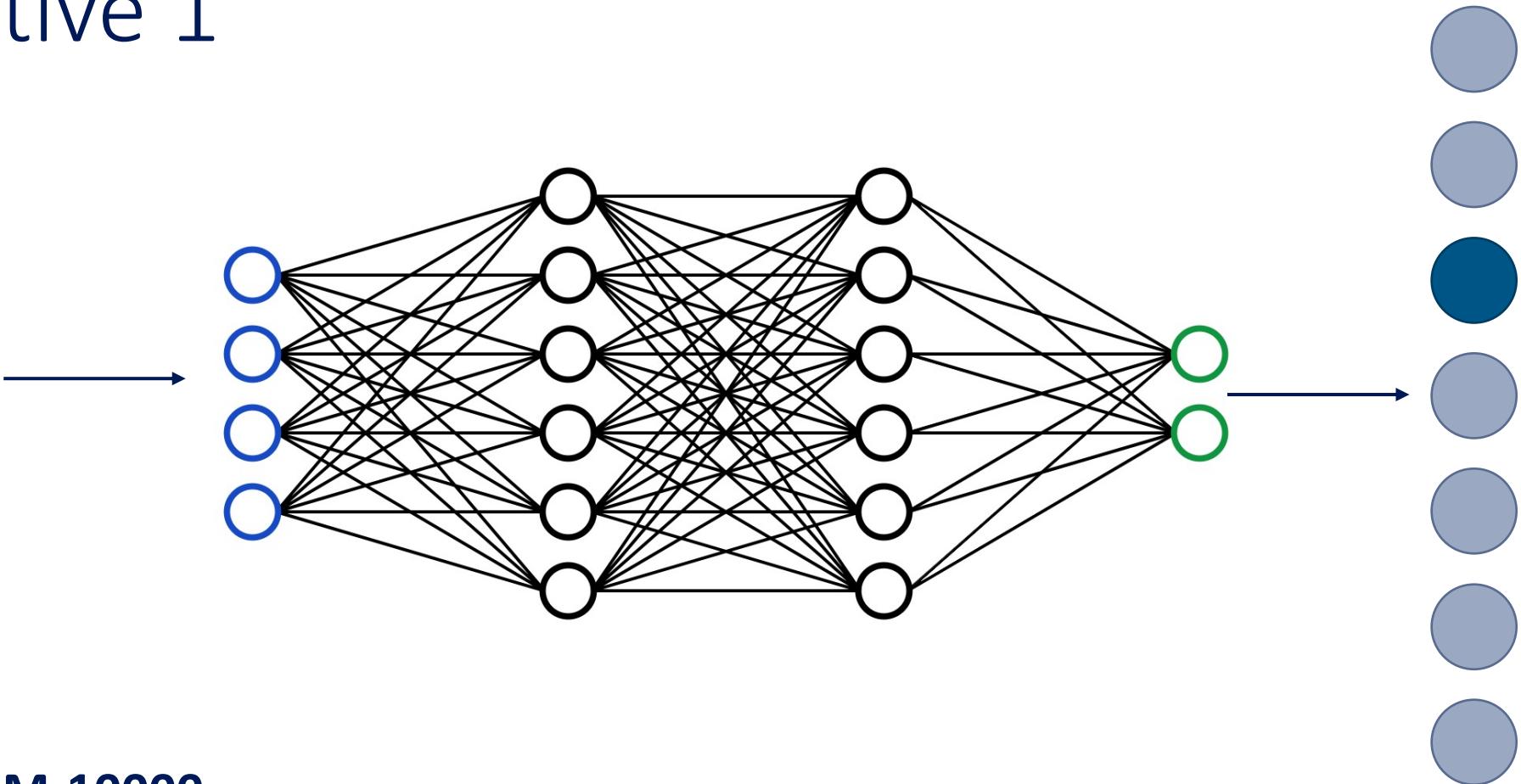
Dataset

<i>Skin Pathology</i>	<i>Number of Images</i>
Actinic keratosis (akiec)	327
Basal cell carcinoma (bcc)	514
Dermatofibroma (df)	115
Melanoma (mel)	1113
Nevus (nv)	6705
Pigmented benign keratosis (bkl)	1099
Vascular lesions (vasc)	142
Total	10015

- PRO: Representative of images that could be taken with a smartphone
- CON: Uneven label distribution

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Objective 1



1. Classify HAM-10000

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Objective 2

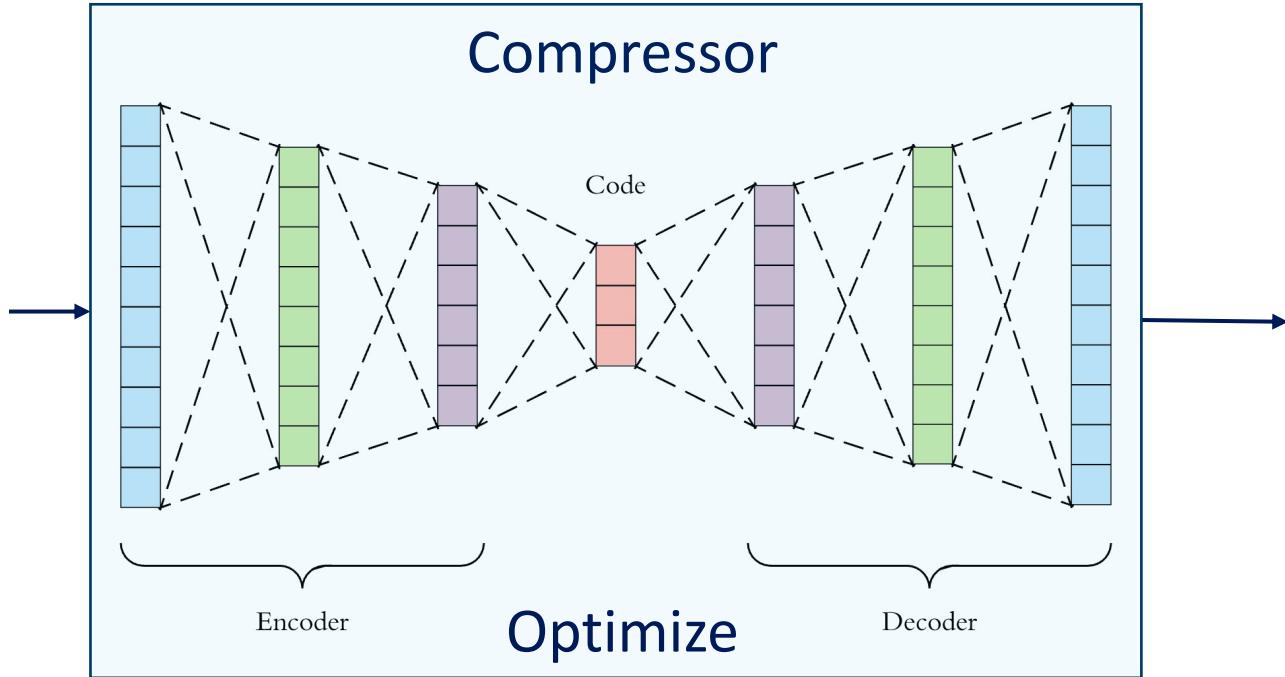
- Current standard image compression is done as JPEG/PNG (etc.)
- Image compression is moving towards Neural Image Compression (**NIC**)

Neural Image Compression for Gigapixel Histopathology Image Analysis

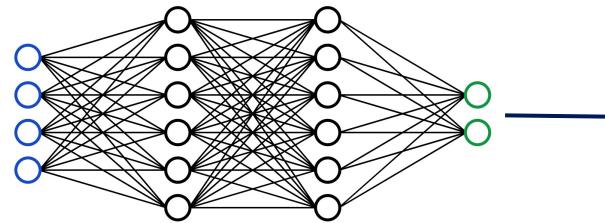
David Tellez, Geert Litjens, Jeroen van der Laak, Francesco Ciompi

Objective 2

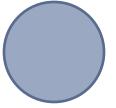
Image



Classify



Score

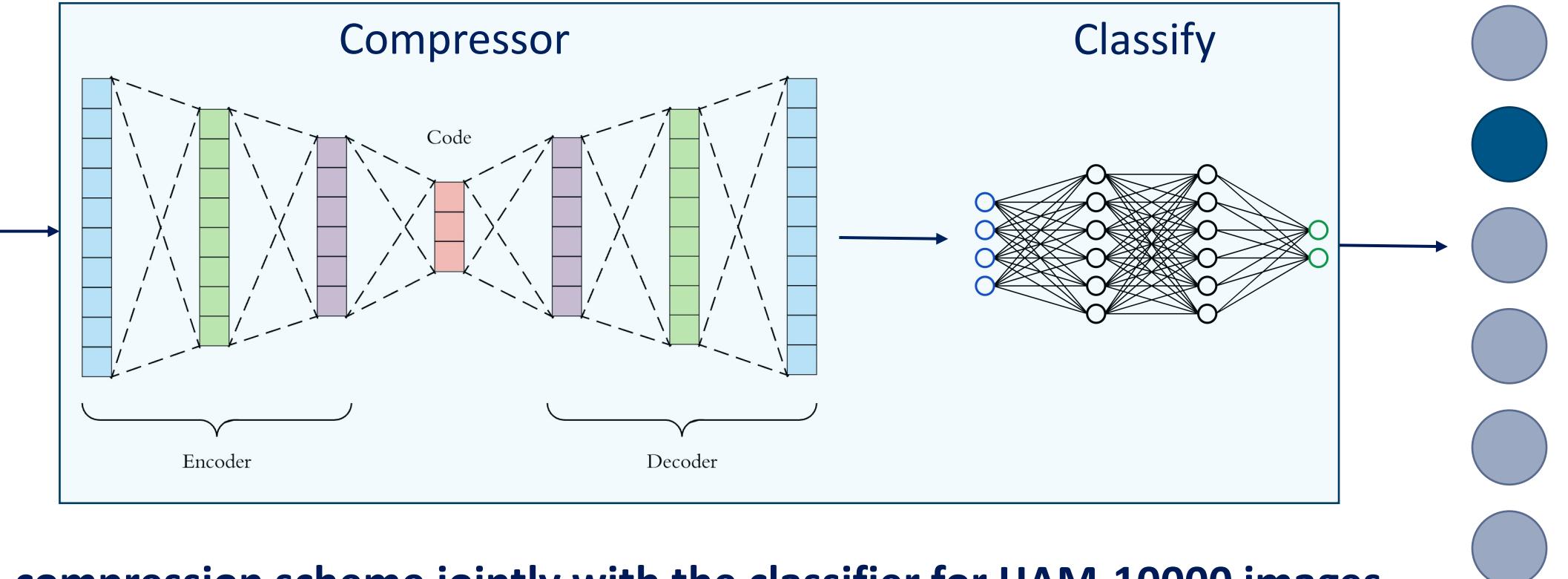


- 2a. Learn a compression scheme for HAM-10000 images
- 2b. See how effective this compression is by analysis using the classifier

Objective 3

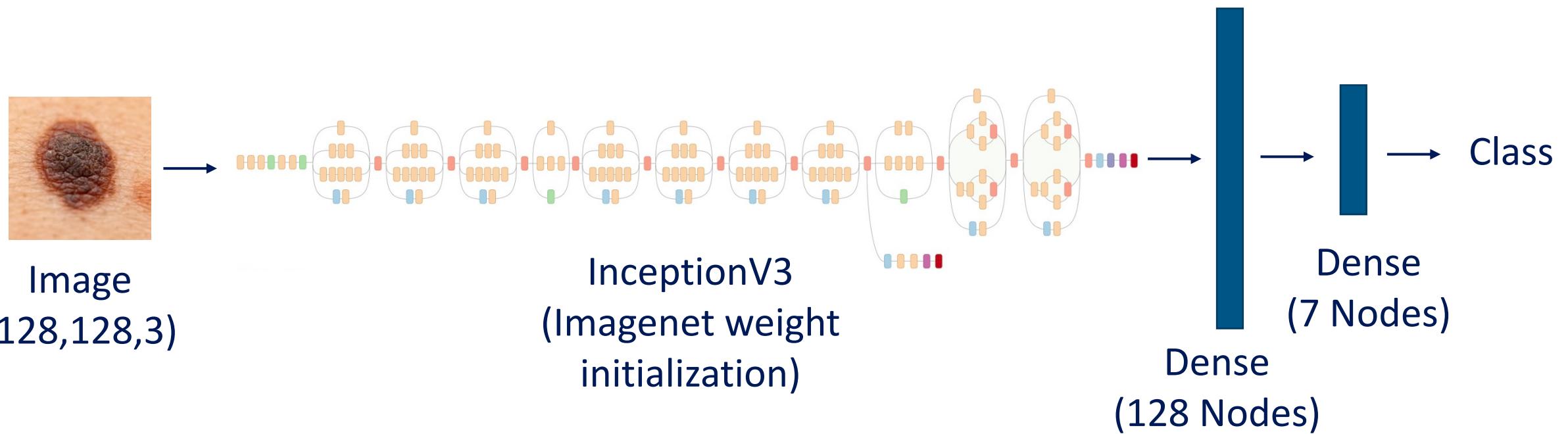
Optimize Together

Image



- 2a. Learn a compression scheme jointly with the classifier for HAM-10000 images
- 2b. After training, extract images and observe “optimal image” for classifier

Classifier Architecture

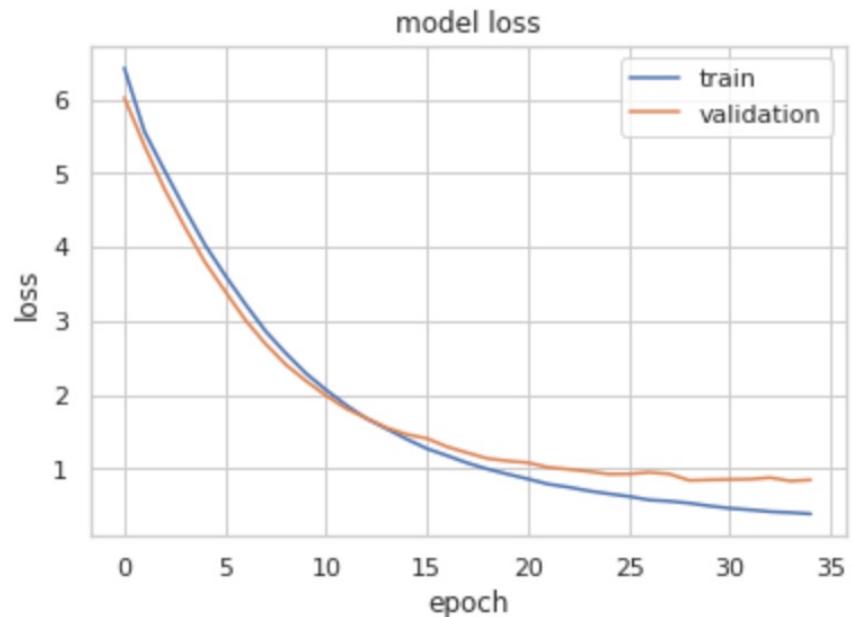
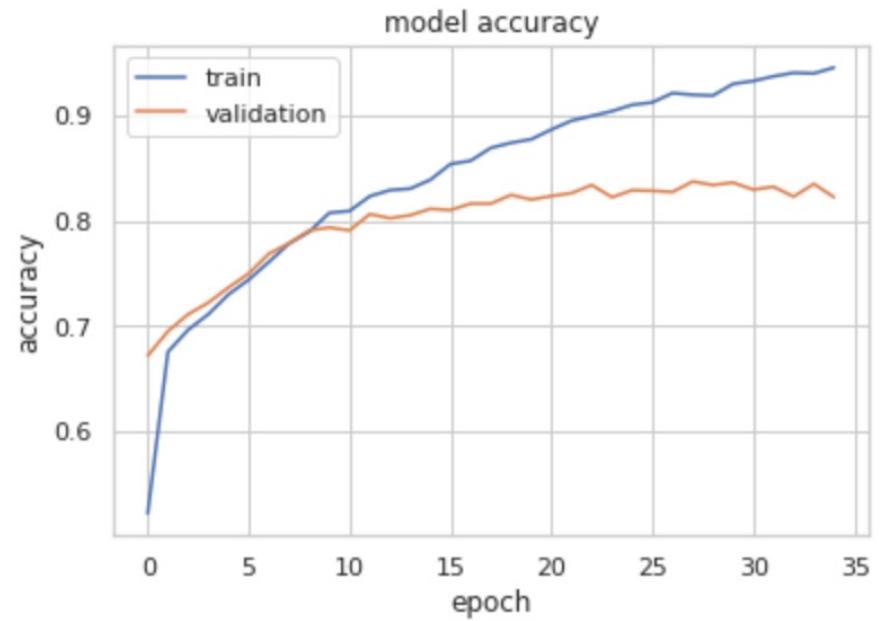


**Dermatologist-level classification of skin cancer with
deep neural networks**

Andre Esteva [✉](#), Brett Kuprel [✉](#), Roberto A. Novoa [✉](#), Justin Ko, Susan M. Swetter, Helen M. Blau &
Sebastian Thrun [✉](#)

Details

- 70/20/10 Train/Validation/Test Split
- Adam optimizer
- L2 regularization
- Data Augmentation
- Categorical Crossentropy Loss
- Normalize inputs to 0-1
- Trained for 35 Epochs



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Classifier Performance

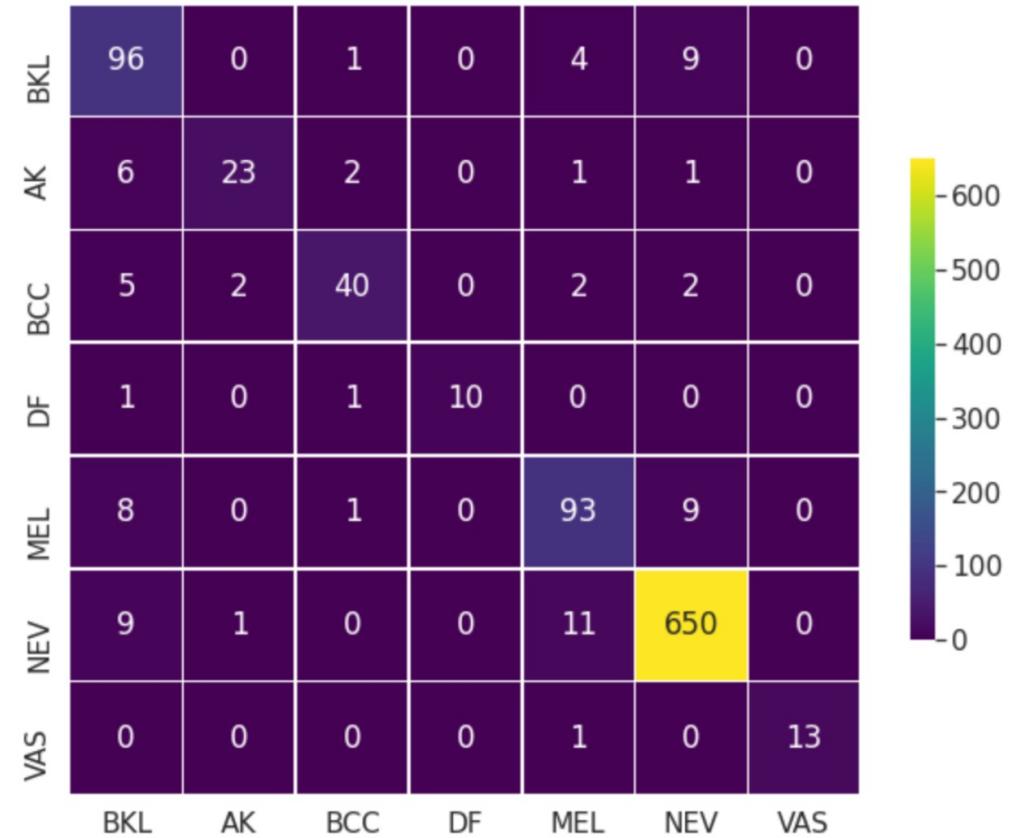
Table 1: Classifier Metrics

Data Set	Loss	Accuracy
Training	0.3819	0.9506
Validation	0.8479	0.8225
Test	0.473367	0.923154

Classifier Performance

Table 2: Classifier Dataset Precision, Recall, F1-Score, and Support

Skin Lesion	Precision	Recall	F1-Score	Support
BKL	0.61	0.69	0.65	110
AK	0.67	0.42	0.52	33
BCC	0.89	0.65	0.75	51
DF	0.70	0.58	0.64.	12
MEL	0.55	0.56	0.56	111
NEV	0.91	0.92	0.92	671
VAS	0.85	0.79	0.81	14
Accuracy			0.82	1002
Macro Average	0.74	0.66	0.69	1002
Weight Average	0.82	0.82	0.82	1002

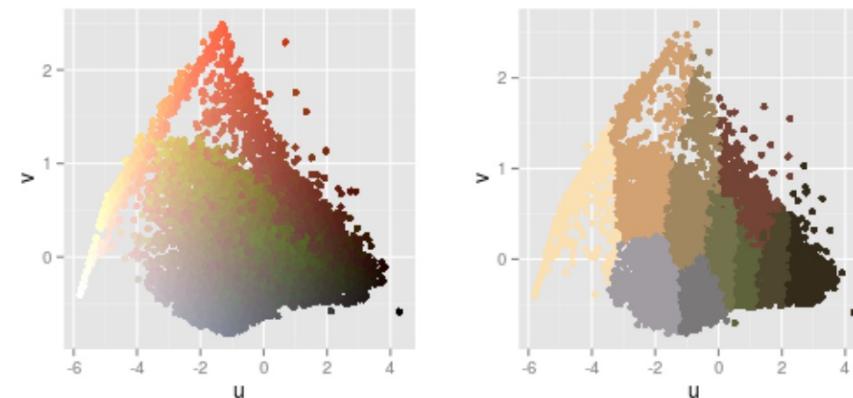


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Good at NEV, Bad at MEL, BKL

Compression Method

- Naïve, first attempt: K-Means Clustering

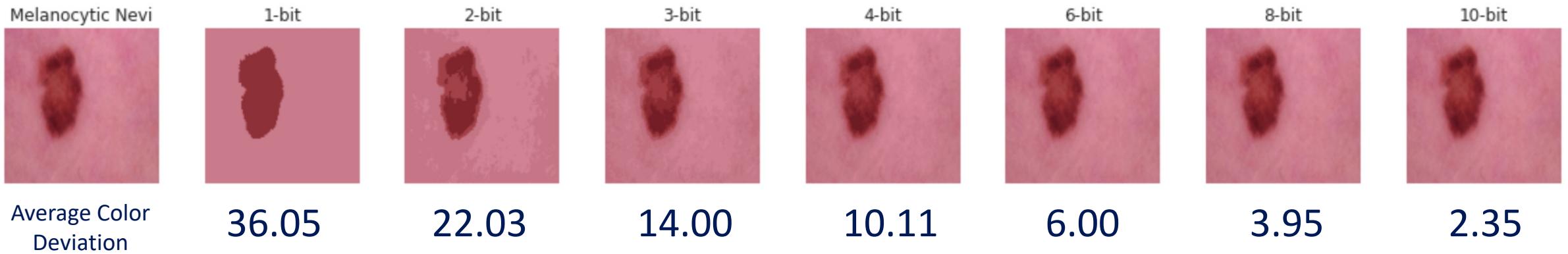


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Carlo Tomasi, Duke CS371

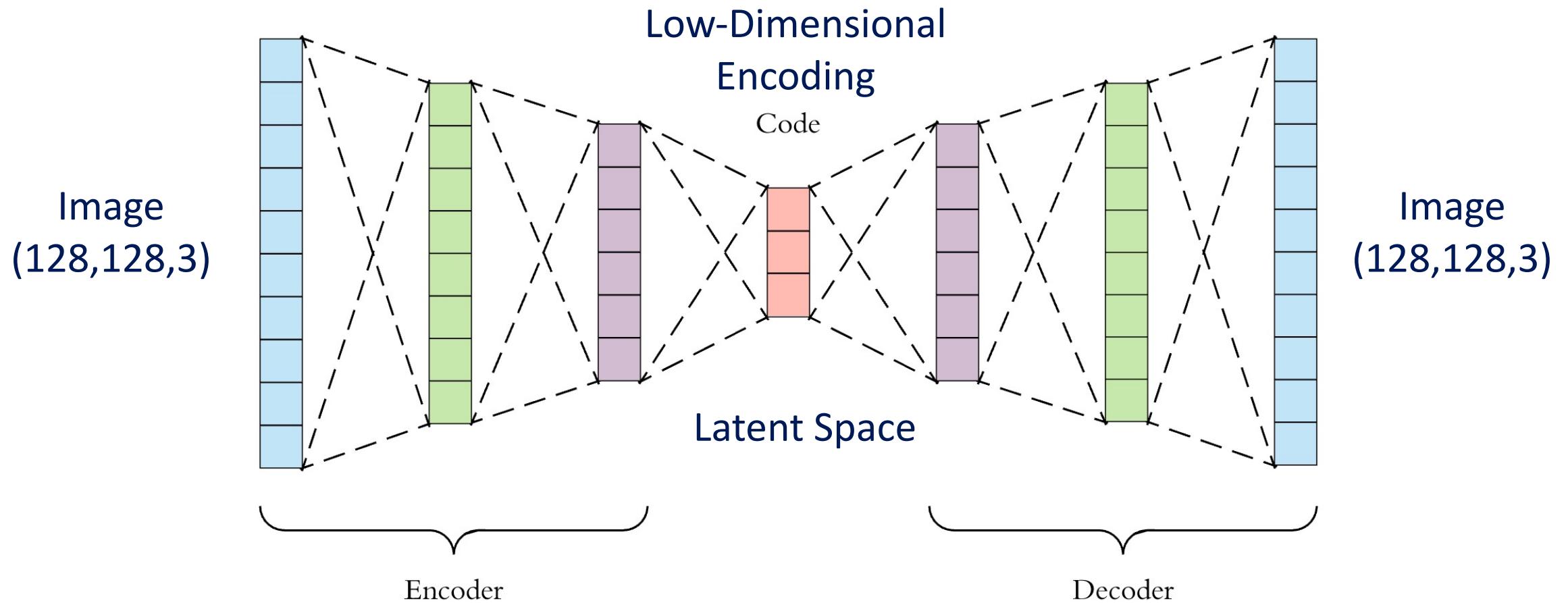
Results

- Key Issue: Not jointly trainable with the classifier.



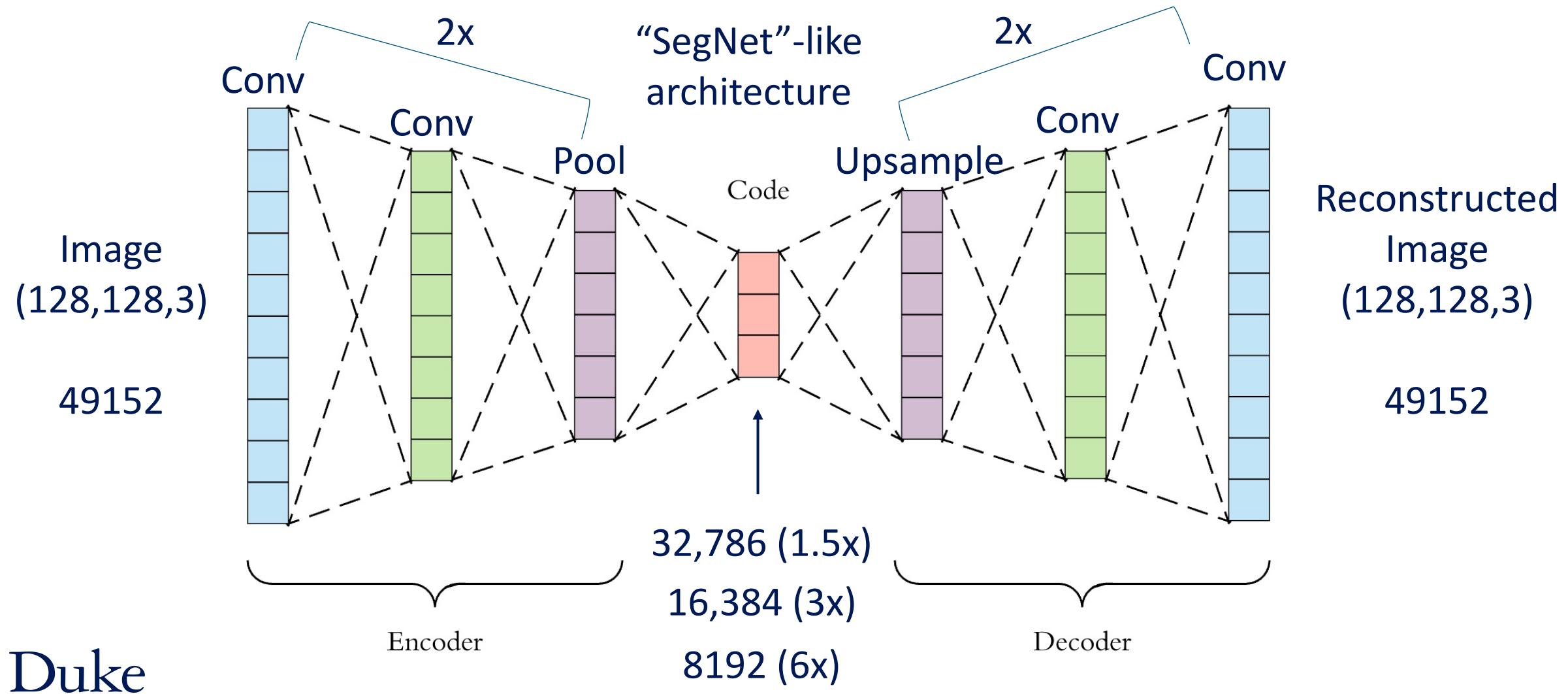
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Autoencoder



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Autoencoder Architecture

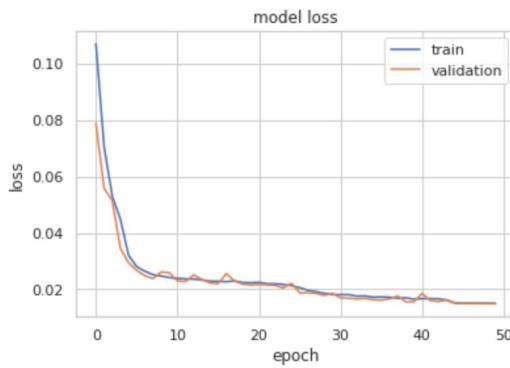


Training Details

- Train for 50 epochs
- Mean Absolute Error loss

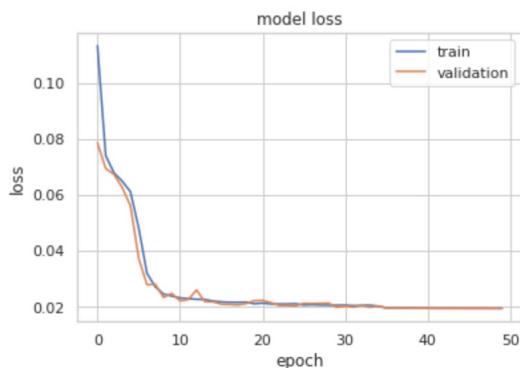
Data Set	Loss	Accuracy
Training	0.0149	-
Validation	0.0149	-
Test	1.1768	0.7538

Table 4: 6X Autoencoder



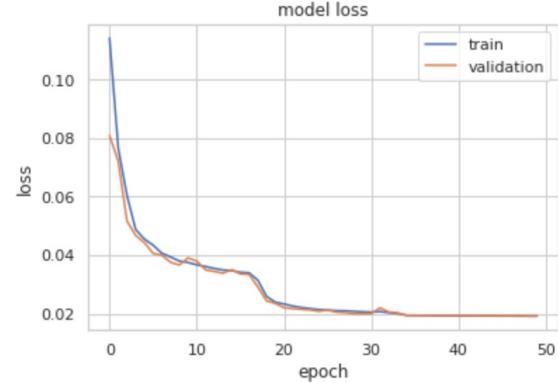
Data Set	Loss	Accuracy
Training	0.0194	-
Validation	0.0194	-
Test	1.2854	0.7315

Table 5: 3X Autoencoder

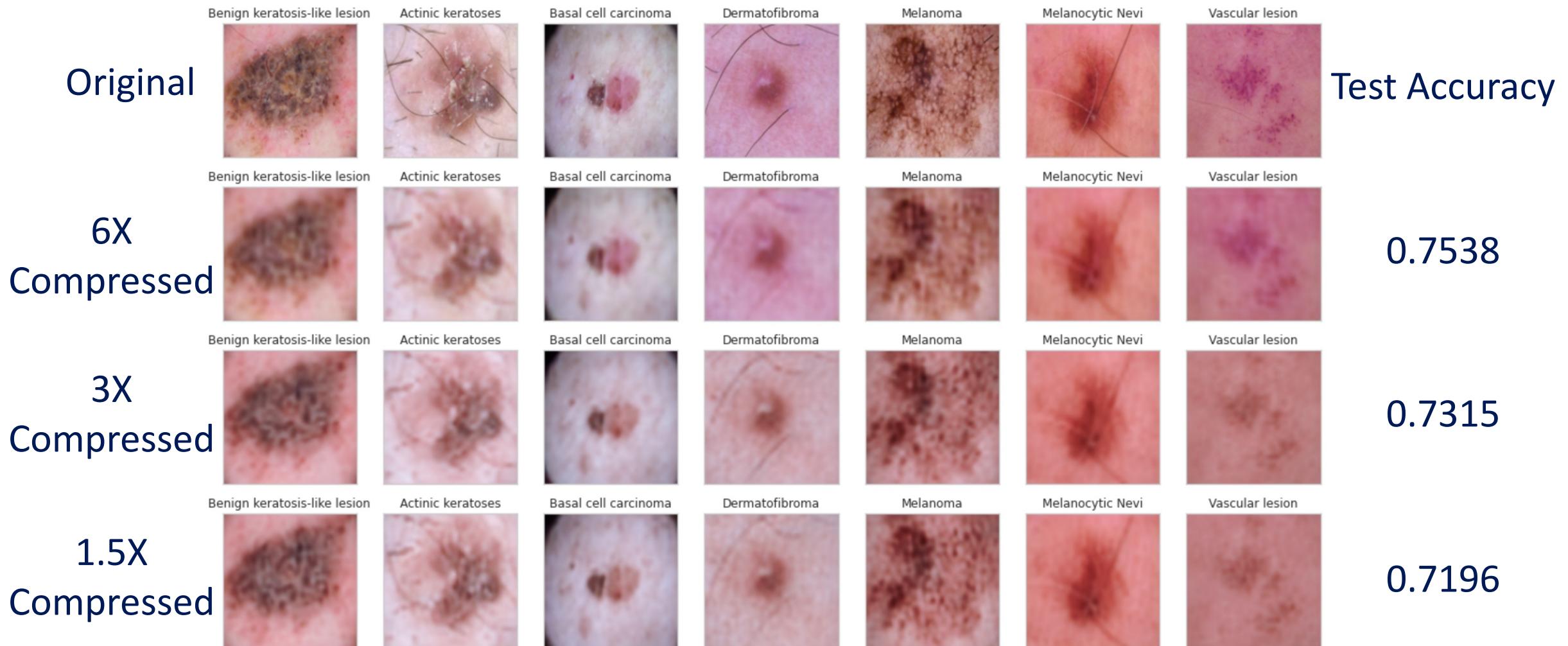


Data Set	Loss	Accuracy
Training	0.0193	-
Validation	0.0193	-
Test	1.3058	0.7196

Table 6: 1.5X Autoencoder



Autoencoder Results

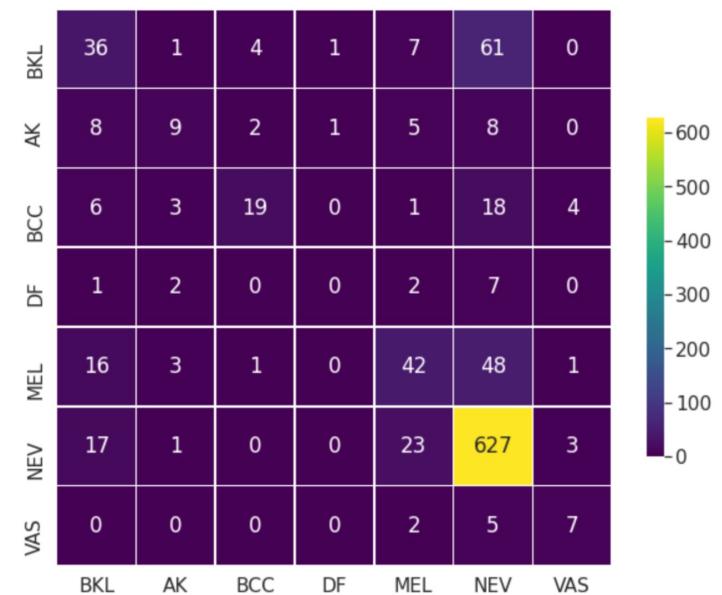


Autoencoder Results – 6X

Table 10: 6X Autoencoder + Pretrained Classifier Precision, Recall, F1-Score, and Support

Skin Lesion	Precision	Recall	F1-Score	Support
BKL	0.78	0.26	0.39	110
AK	0.57	0.24	0.34	33
BCC	0.44	0.78	0.57	51
DF	0.00	0.00	0.00	12
MEL	0.52	0.32	0.39	111
NEV	0.81	0.96	0.88	671
VAS	1.00	0.07	0.13	14
Accuracy			0.76	1002
Macro Average	0.59	0.38	0.39	1002
Weight Average	0.74	0.76	0.72	1002

Figure 7: 6X Autoencoder + Pretrained Classifier Confusion Matrix

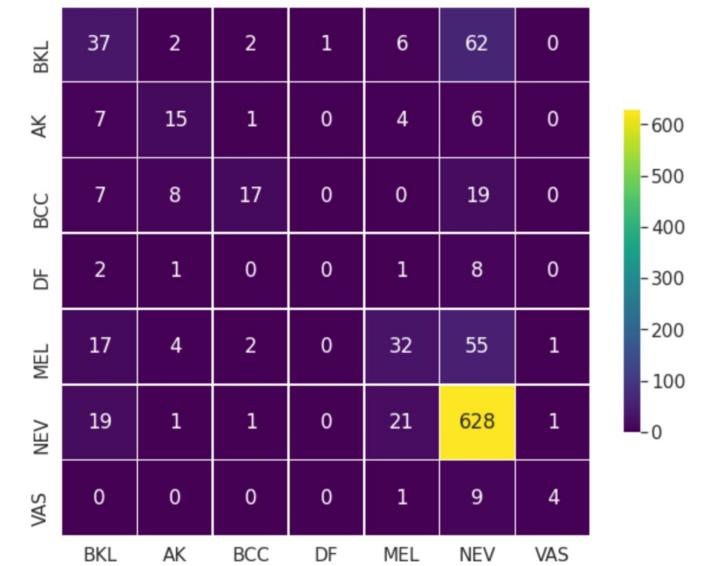


Autoencoder Results -3X

Table 11: 3X Autoencoder + Pretrained Classifier Precision, Recall, F1-Score, and Support

Skin Lesion	Precision	Recall	F1-Score	Support
BKL	0.60	0.60	0.60	110
AK	0.67	0.48	0.56	33
BCC	0.63	0.78	0.70	51
DF	0.00	0.00	0.00	12
MEL	0.56	0.48	0.52	111
NEV	0.91	0.94	0.92	671
VAS	1.00	0.93	0.96	14
Accuracy			0.82	1002
Macro Average	0.62	0.60	0.61	1002
Weight Average	0.80	0.82	0.81	1002

Figure 8: 3X Autoencoder + Pretrained Classifier Classifier Confusion Matrix

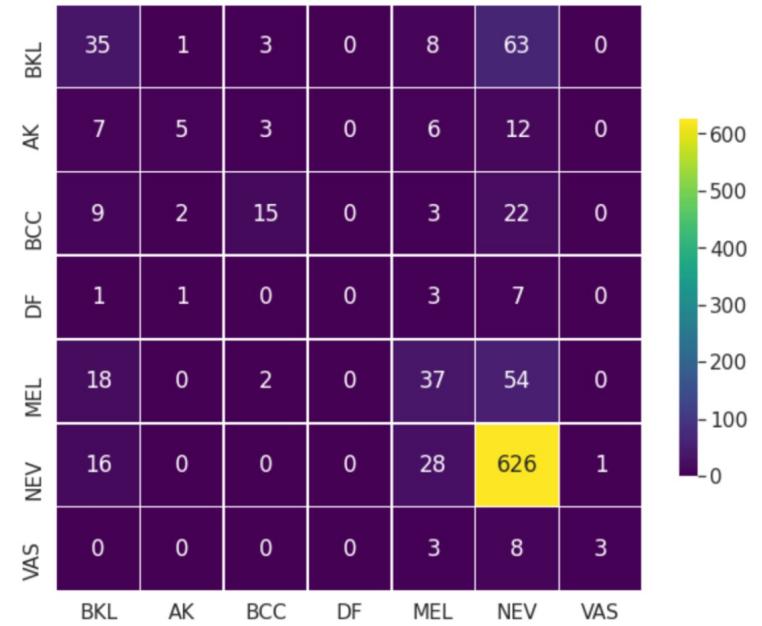


Autoencoder Results -1.5X

Table 12: 1.5X Autoencoder + Pretrained Classifier Precision, Recall, F1-Score, and Support

Skin Lesion	Precision	Recall	F1-Score	Support
BKL	0.60	0.51	0.55	110
AK	0.52	0.33	0.41	33
BCC	0.56	0.61	0.58	51
DF	0.00	0.00	0.00	12
MEL	0.53	0.44	0.48	111
NEV	0.86	0.94	0.90	671
VAS	0.70	0.50	0.58	14
Accuracy			0.78	1002
Macro Average	0.54	0.48	0.50	1002
Weight Average	0.76	0.78	0.77	1002

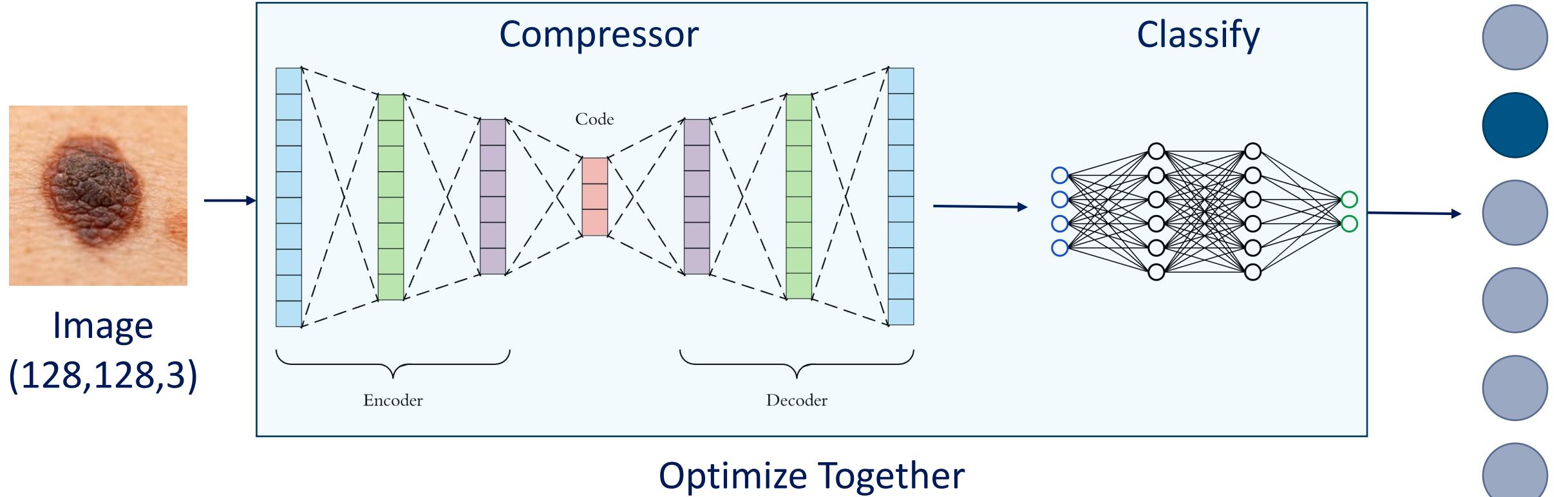
Figure 9: 1.5X Autoencoder + Pretrained Classifier Confusion Matrix



Autoencoder Results Summary

- All three autoencoders had similar accuracies, with the 6X being the most accurate and 1.5X being the least accurate on test set
- All three models did the best at classifying NEV
- All three models were unable to properly classify DF

Joint Training Architecture



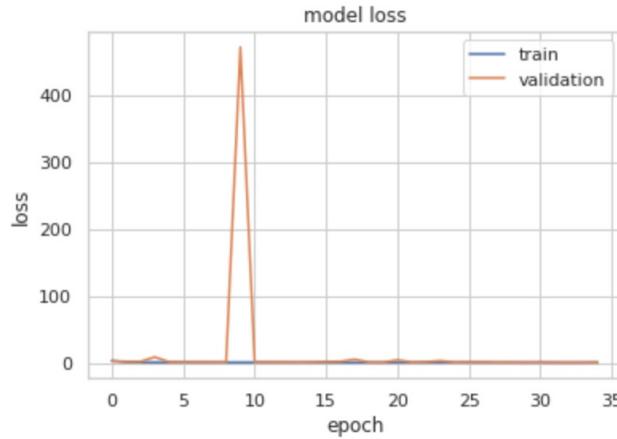
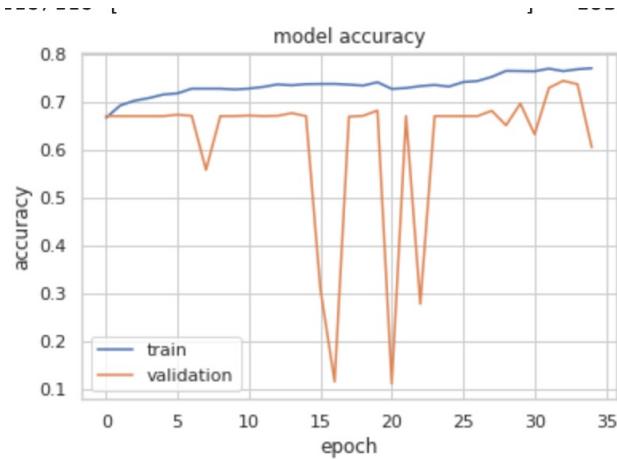
Duke

Training Details

- 70/20/10 Train/Validation/Test Split
- Adam optimizer
- L2 regularization
- Data Augmentation
- Categorical Crossentropy Loss
- Normalize inputs to 0-1
- Trained for 35 Epochs
- Checkpointing due to unstable validation accuracy

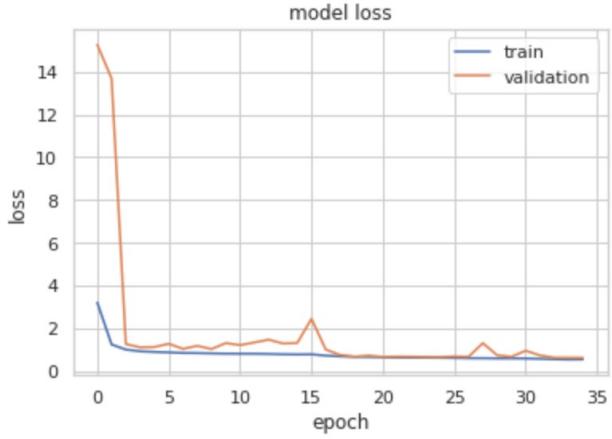
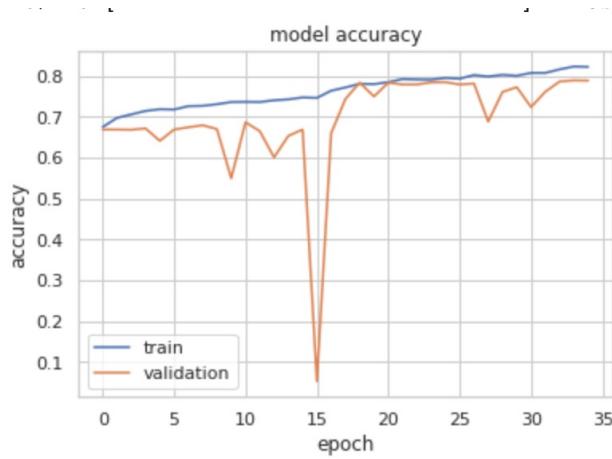
Duke

Training Details

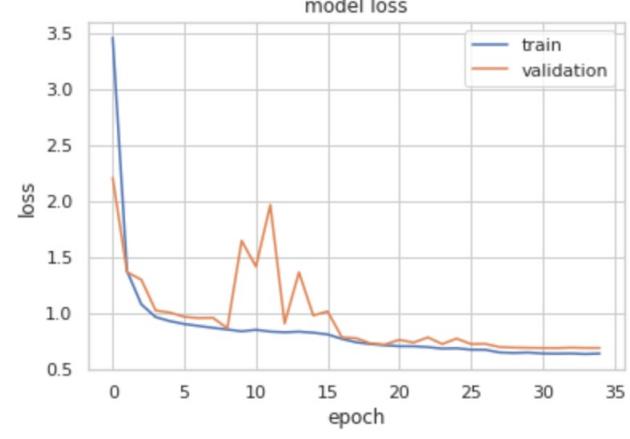
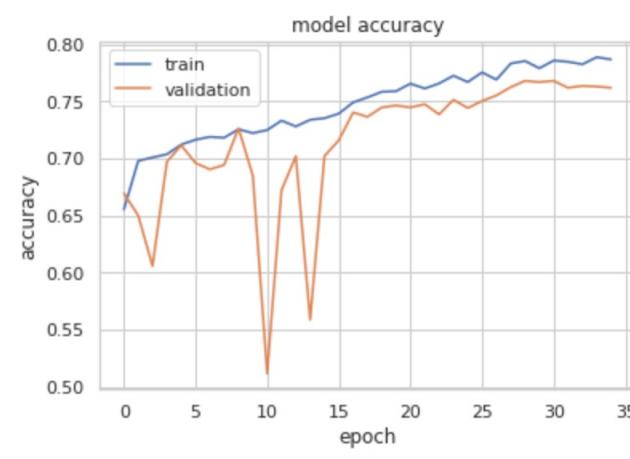


1.5X

Duke

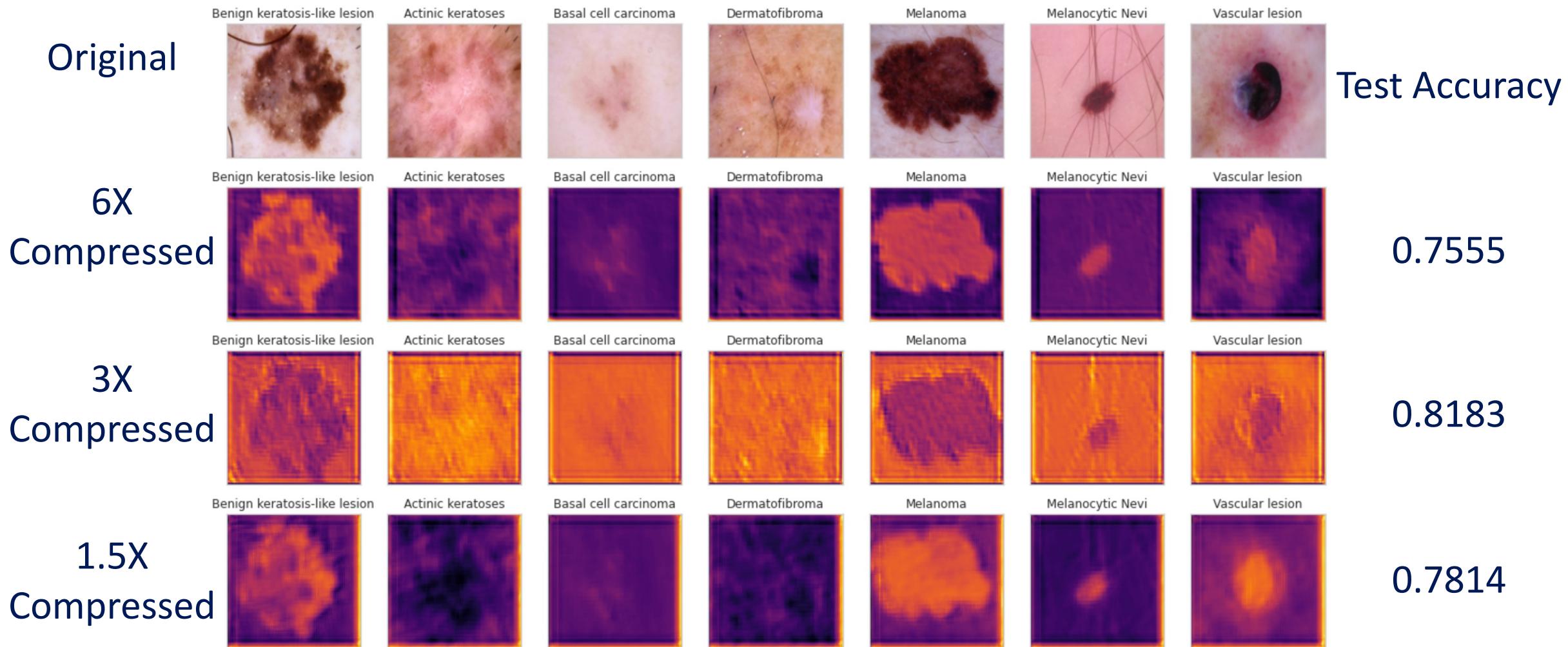


3X



6X

Joint Training Results



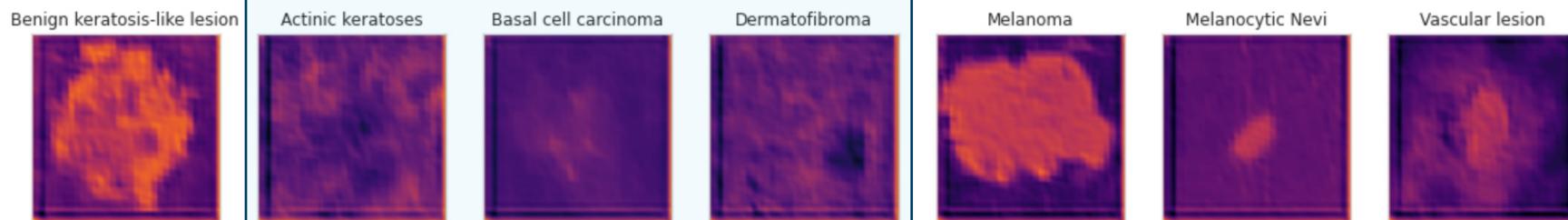
Contrast is critical for classification!

Joint Training Results

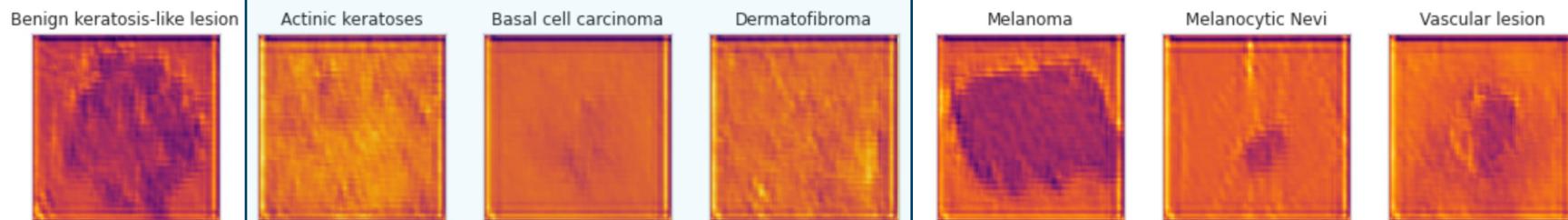
Original



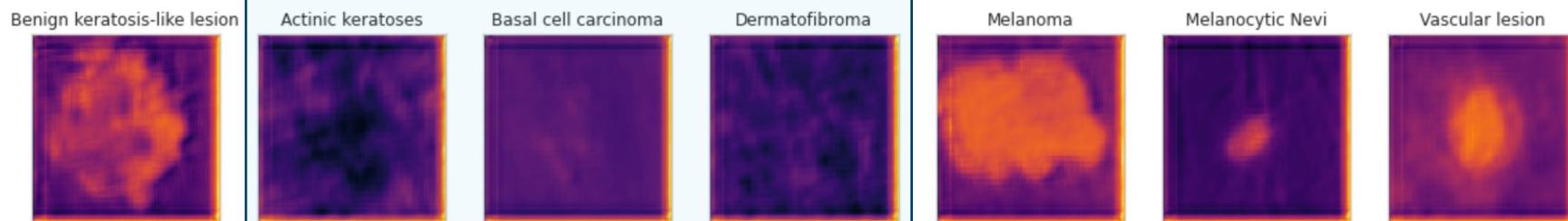
6X



3X



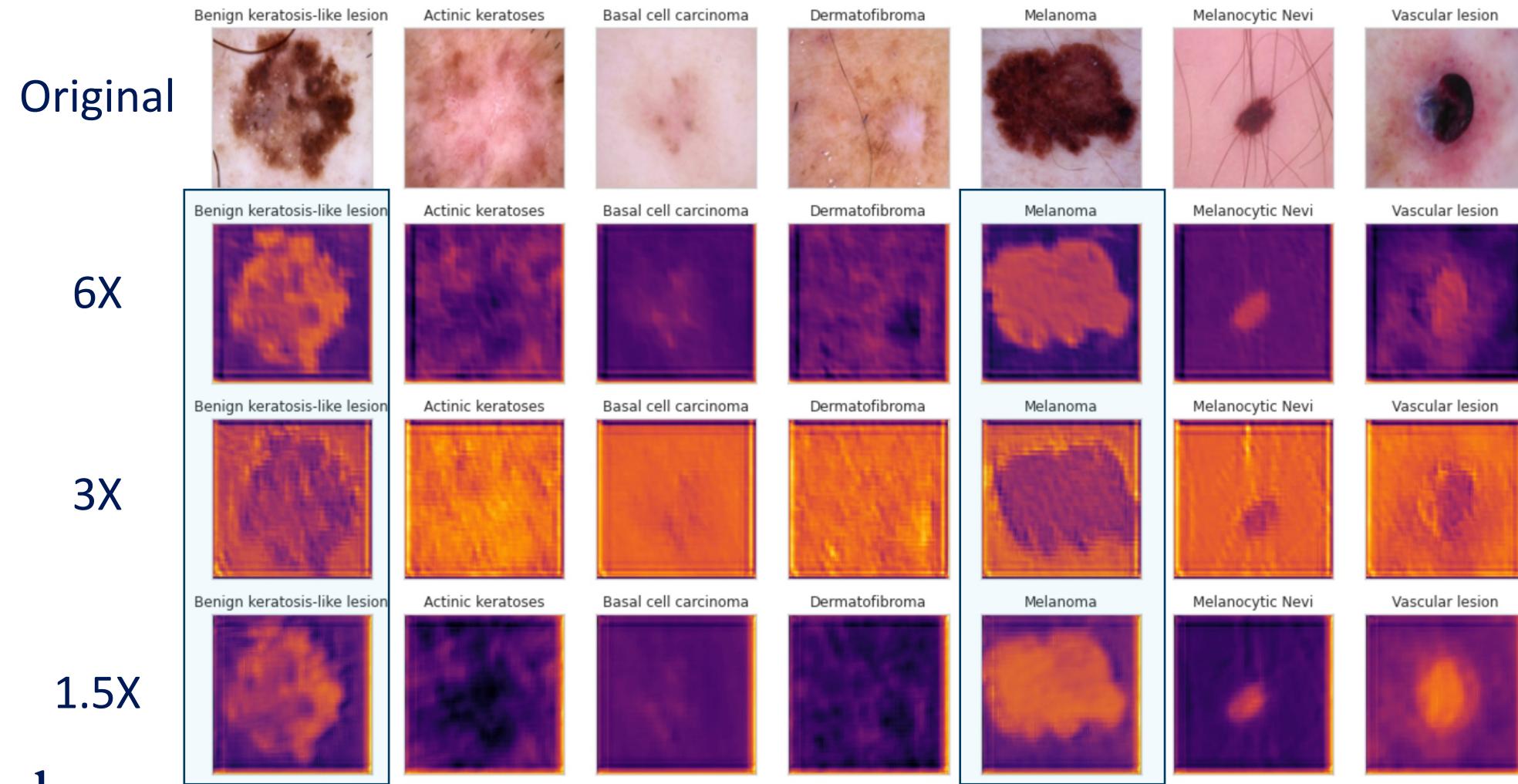
1.5X



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DF, BCC, AK have poor contrast, may suggest why model is confused.

Joint Training Results



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BKL and MEL have similar contrast, theory for confusion matrix results

Joint Training Results

Data Set	Loss	Accuracy
Training	0.6933	0.7699
Validation	0.7685	0.7432
Test	0.7198	0.7555

Table 7: 6X Joint Model

Data Set	Loss	Accuracy
Training	0.5469	0.8261
Validation	0.6289	0.7898
Test	0.5478	0.8183

Table 8: 3X Joint Model

Data Set	Loss	Accuracy
Training	0.6291	0.7869
Validation	0.6870	0.7621
Test	0.6098	0.7814

Table 9: 1.5X Joint Model

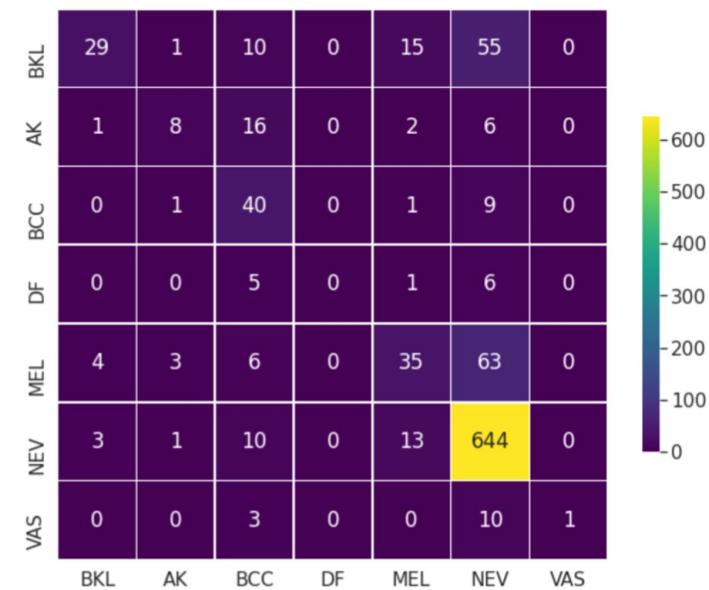
3X Encoder Architecture had the best performance

Joint Training Results – 6X

Table 13: 6X Compression Jointly Trained Model Precision, Recall, F1-Score, and Support

Skin Lesion	Precision	Recall	F1-Score	Support
BKL	0.78	0.26	0.39	110
AK	0.57	0.24	0.34	33
BCC	0.44	0.78	0.57	51
DF	0.00	0.00	0.00	12
MEL	0.52	0.32	0.39	111
NEV	0.81	0.96	0.88	671
VAS	1.00	0.07	0.13	14
Accuracy			0.76	1002
Macro Average	0.59	0.38	0.39	1002
Weight Average	0.74	0.76	0.72	1002

Figure 10: 6X Compression Jointly Trained Model Classifier Confusion Matrix

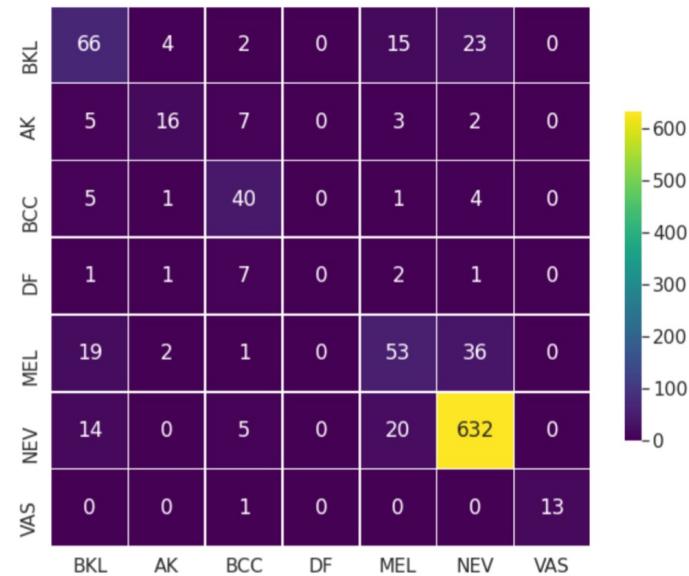


Joint Training Results – 3X

Table 14: 3X Compression Jointly Trained Model Precision, Recall, F1-Score, and Support

Skin Lesion	Precision	Recall	F1-Score	Support
BKL	0.60	0.60	0.60	110
AK	0.67	0.48	0.56	33
BCC	0.63	0.78	0.70	51
DF	0.00	0.00	0.00	12
MEL	0.56	0.48	0.52	111
NEV	0.91	0.94	0.92	671
VAS	1.00	0.93	0.96	14
Accuracy			0.82	1002
Macro Average	0.62	0.60	0.61	1002
Weight Average	0.80	0.82	0.81	1002

Figure 11: 3X Compression Jointly Trained Model Classifier Confusion Matrix

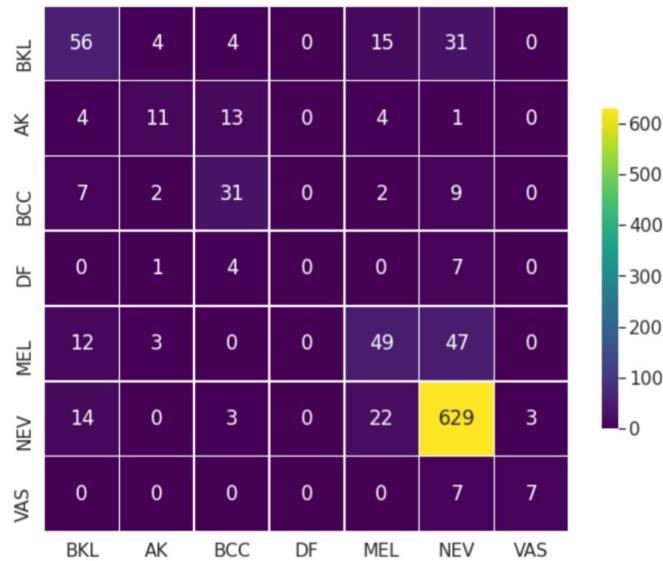


Joint Training Results -1.5X

Table 15: 1.5X Compression Jointly Trained Model Precision, Recall, F1-Score, and Support

Skin Lesion	Precision	Recall	F1-Score	Support
BKL	0.60	0.51	0.55	110
AK	0.52	0.33	0.41	33
BCC	0.56	0.61	0.58	51
DF	0.00	0.00	0.00	12
MEL	0.53	0.44	0.48	111
NEV	0.86	0.94	0.90	671
VAS	0.70	0.50	0.58	14
Accuracy			0.78	1002
Macro Average	0.54	0.48	0.50	1002
Weight Average	0.76	0.78	0.77	1002

Figure 12: 1.5X Compression Jointly Trained Model Classifier Confusion Matrix



Joint Training Results Summary

- 3X model had **superior average precision and recall** to 1.5X and 6X models.
 - Follows trend seen with autoencoder training
- 3X Model **was able to distinguish vascular lesions from nevi**, other two models were not
- 3X Model was **superior at classifying BKL**
- All three models **confused between BCC, AK, DF**
- 3X Model test accuracy **10% less accurate than classifier alone (82%)**
- All three models performed the **best on nevus classification**

Takeaways

- **Image background contrast is critical for HAM-10000 classification**
 - Imaging scheme: highlight contrast between lesion and background
 - Compression scheme: possible to only encode boundaries between lesion and background
- **Latent space size may not directly be correlated to compression effectiveness**
- **Autoencoder effectiveness may not be directly correlated to classification effectiveness**

Future Steps

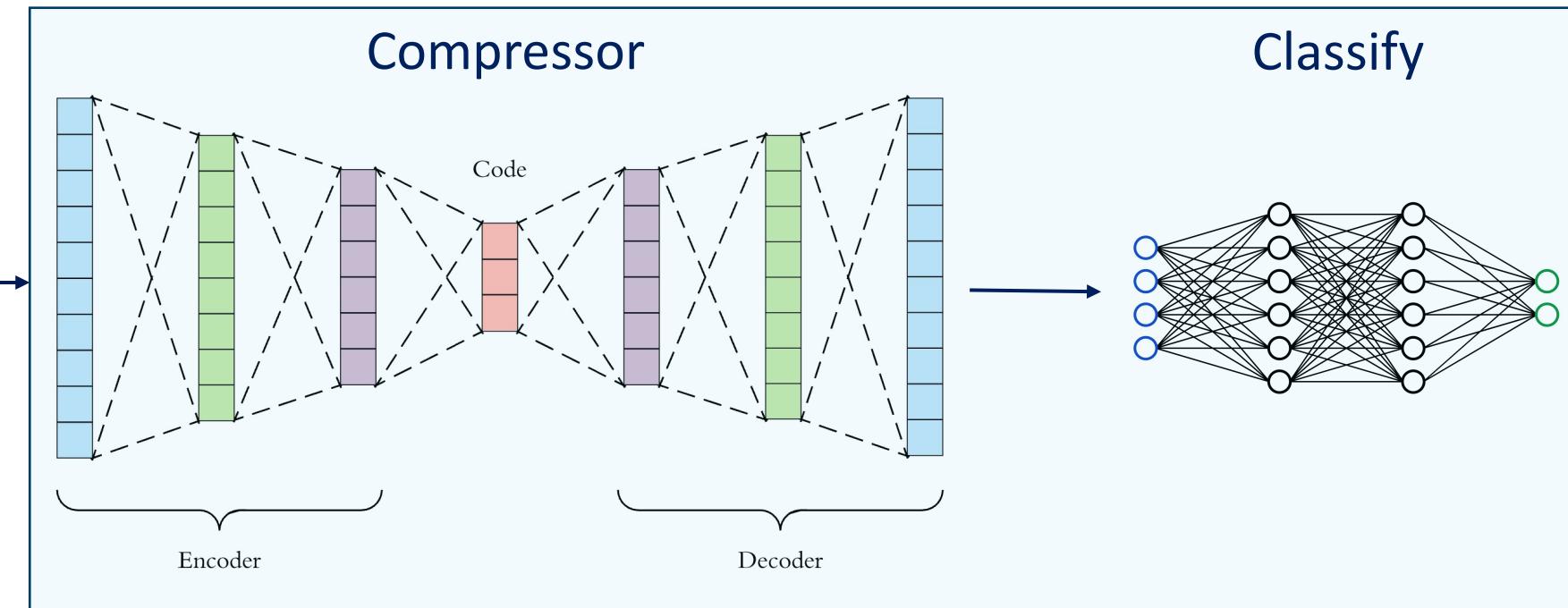
- Train physical layer that serves as an edge detection filter, add to front of joint model and observe performance.
- Address unbalanced dataset issue
- Test different underlying classification architectures

Thank you!

- Any questions?



Image
(128,128,3)



Duke

Score

