

# MACHINE-LEARNING CLASSIFICATION OF BLOWING SNOW FROM WEBCAM IMAGERY

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June 6<sup>th</sup>, 2020

19<sup>th</sup> Conference on Mountain Meteorology



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# OVERVIEW

- CSTAR overview
- Convolutional neural networks overview
- My research goals
- First round of training – binary model
- Second round of training – classification model
- Meteorological characteristics of each category
- Future work

## CSTAR INTRO

- NOAA/NWS Collaborative Science Technology, and Applied Research
- Aims to improve hazardous winter weather forecasting
- Wyoming impacted by several types of hazardous winter wx
  - High winds
  - Snow squalls
  - **Blowing snow (BLSN)**
- Focus on HRRR (High-Resolution Rapid Refresh) model
  - Validation of HRRR forecasts
  - HRRR-based predictions
  - BLSN product

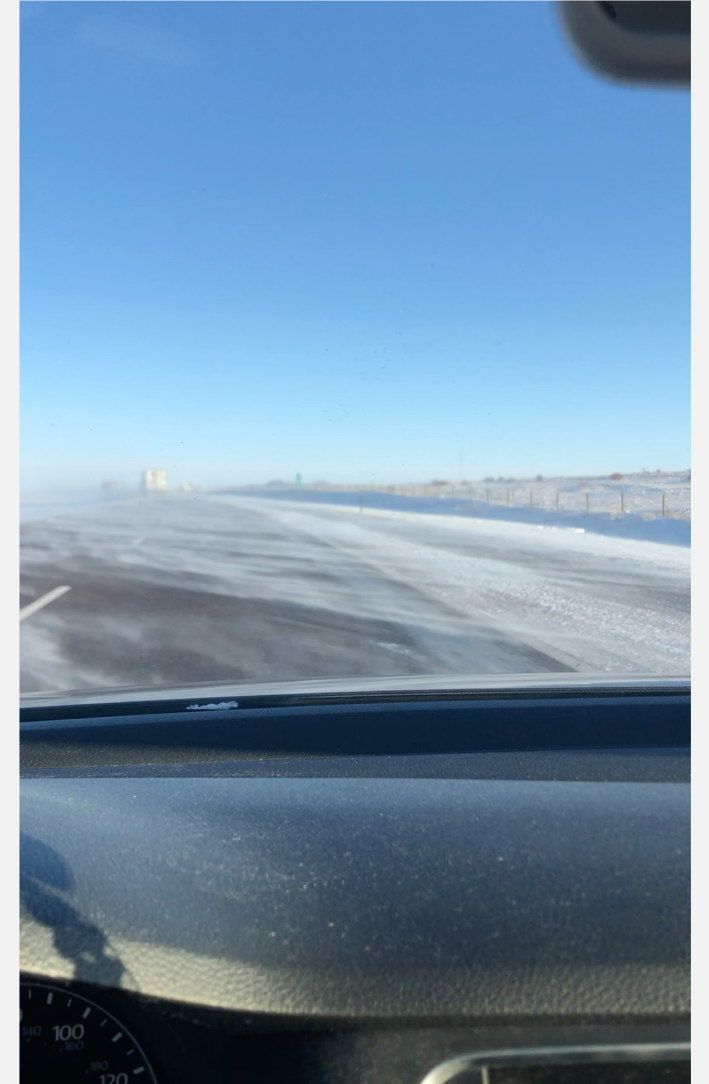
# BLOWING SNOW RISKS

Blowing snow affects travel

- Low visibility
- Hard to predict

Depends on:

- Snow cover
- Snow age
- Other meteorological variables



# U.S. Annual Hours of Blowing Snow

Estimated from Modeled Data

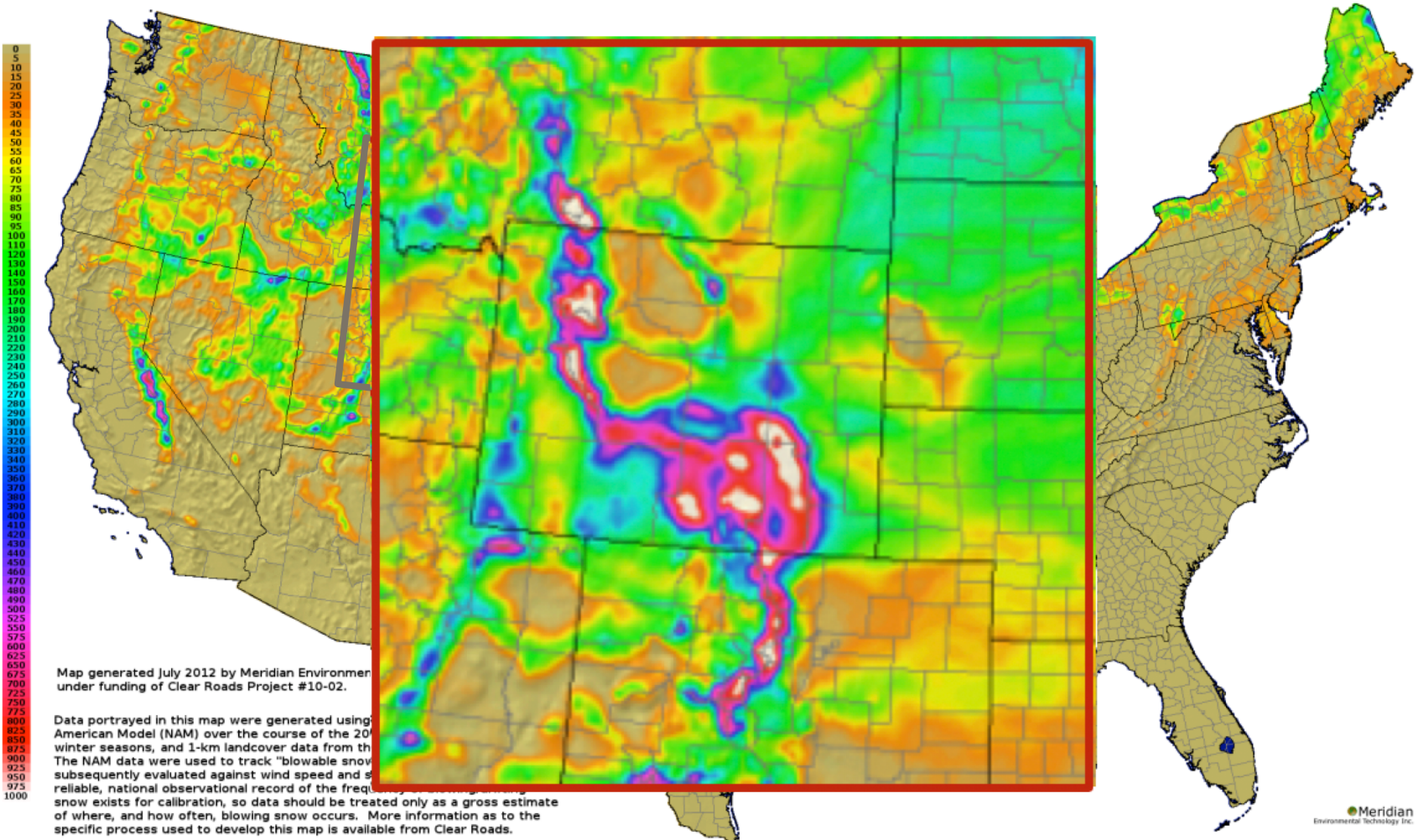


Figure 4: Map of Average Annual Duration of Blowing/Drifting Snow (in hours) as developed under this project.

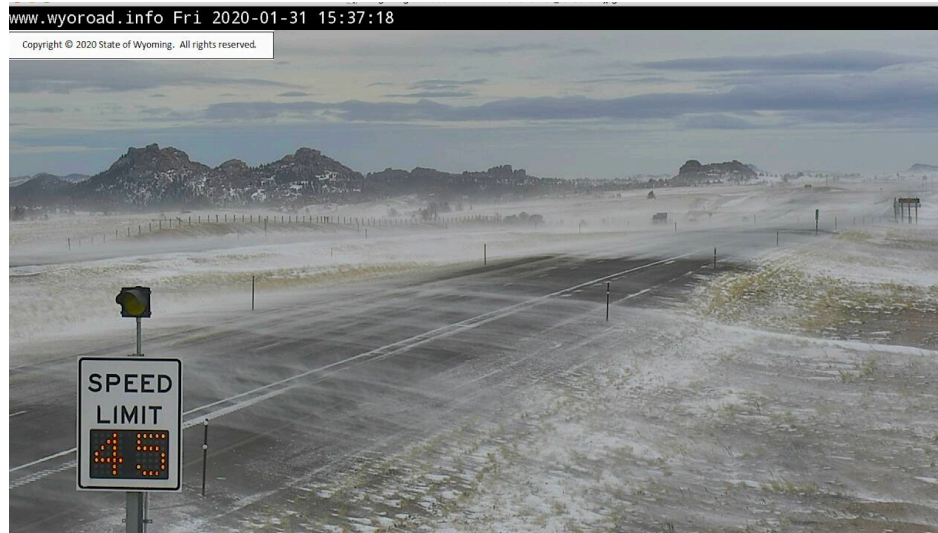
# CONVOLUTIONAL NEURAL NETWORKS

- Image classification uses deep convolutional neural networks (CNNs)
- Defined as “a family of **parametric, non-linear** and **hierarchical representation learning functions**, which are massively optimized...to **encode domain knowledge**, i.e. domain invariances, stationarity.” -- Efstratios Gavves

# CNNS USEFUL FOR IMAGE CLASSIFICATION

- CNNs used for images of blowing snow from webcam images
  - Identifies patterns and features from training dataset
  - Classify based on characteristics only computer sees

What I see



What a computer sees

08	02	22	97	38	15	00	40	00	75	04	05	07	78	52	12	50	77	91	08
49	49	99	40	17	81	18	57	60	87	17	40	98	43	69	48	04	56	62	00
81	49	31	73	55	79	14	29	93	71	40	67	53	88	30	03	49	13	36	65
52	70	95	23	04	60	11	42	69	24	68	56	01	32	56	71	37	02	36	91
22	31	16	71	51	67	63	89	41	92	36	54	22	40	40	28	66	33	13	80
24	47	32	60	99	03	45	02	44	75	33	53	78	36	84	20	35	17	12	50
32	98	81	28	64	23	67	10	26	38	40	67	59	54	70	66	18	38	64	70
67	26	20	68	02	62	12	20	95	63	94	39	63	08	40	91	66	49	94	21
24	55	58	05	66	73	99	26	97	17	78	78	96	83	14	88	34	89	63	72
21	36	23	09	75	00	76	44	20	45	35	14	00	61	33	97	34	31	33	95
78	17	53	28	22	75	31	67	15	94	03	80	04	62	16	14	09	53	56	92
16	39	05	42	96	35	31	47	55	58	88	24	00	17	54	24	36	29	85	57
86	56	00	48	35	71	89	07	05	44	44	37	44	60	21	58	51	54	17	58
19	80	81	68	05	94	47	69	28	73	92	13	86	52	17	77	04	89	55	40
04	52	08	83	97	35	99	16	07	97	57	32	16	26	26	79	33	27	98	66
88	36	68	87	57	62	20	72	03	46	33	67	46	55	12	32	63	93	53	69
04	42	16	73	38	25	39	11	24	94	72	18	08	46	29	32	40	62	76	36
20	69	36	41	72	30	23	88	34	62	99	69	82	67	59	85	74	04	36	16
20	73	35	29	78	31	90	01	74	31	49	71	48	86	81	16	23	57	05	54
01	70	54	71	83	51	54	69	16	92	33	48	61	43	52	01	89	19	67	48

# CNNs CONVOLVE OVER THEIR INPUT

3.0	3.0	3.0
3.0	3.0	3.0
3.0	2.0	3.0

Max numbers being retained – most important features are carried on through the model (pooling)

Input – image RGB values

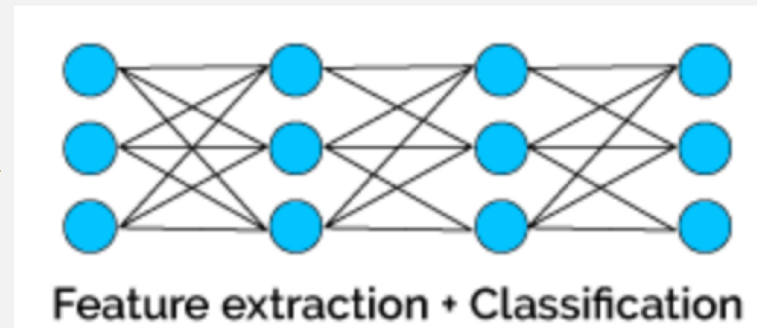
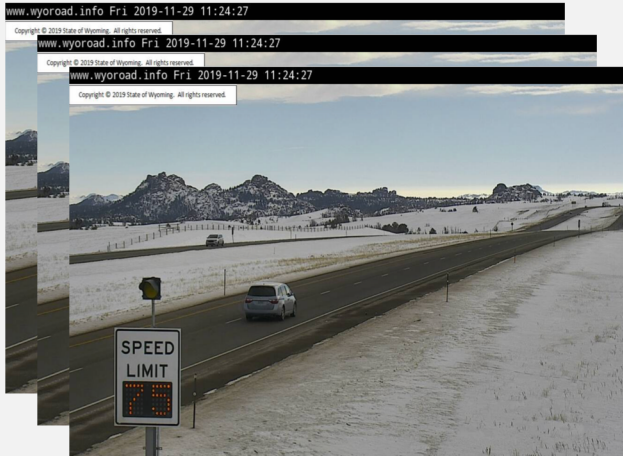
3	3	2	1	0
0	0	1	3	1
3	1	2	2	3
2	0	0	2	2
2	0	0	0	1

Shaded: Sliding convolution kernel with size 3x3



# GOALS OF THE BLOWING SNOW ML MODEL: DEEP LEARNING ALGORITHM OVERVIEW

Input:  
real-time webcam images



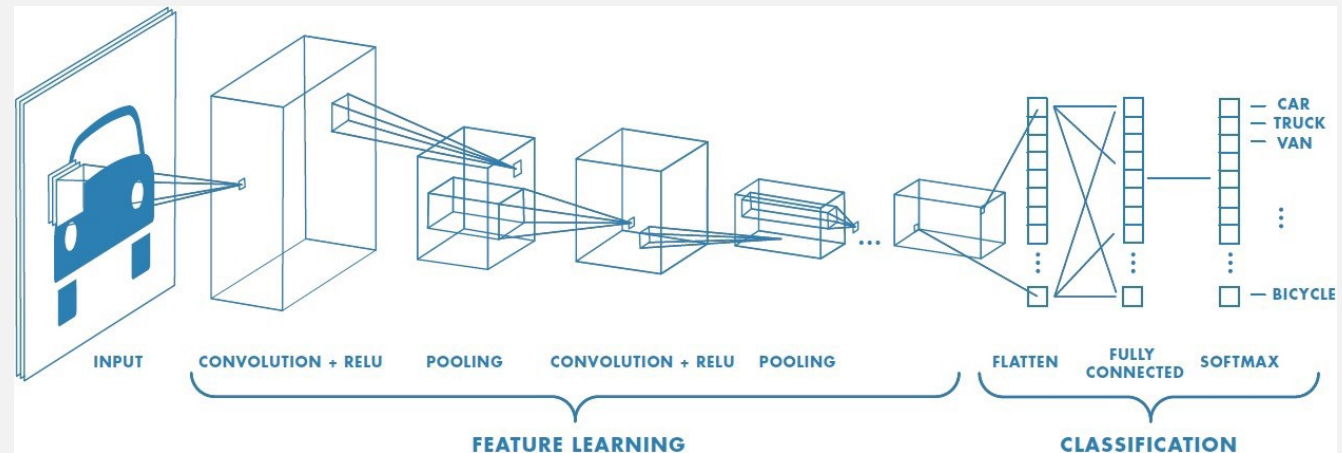
Output:  
Is there blowing  
snow in these  
images? No

## GOALS OF THE BLOWING SNOW ML MODEL

- Model accurately classifies blowing snow in real-time upon seeing new images
- Operational use → model notifies forecaster when a webcam detects blowing snow
  - Aid forecasters in blowing snow events with automated webcam detection
- Develop an observational database of BLSN intensity for HRRR-based BLSN product
  - Outputs a probability of a certain intensity of BLSN

# MODEL ARCHITECTURE OVERVIEW

- 3 convolutional layers
  - First layer (input layer) – 32 filters
    - Takes in pixel height, width, **features (RGB values)**
    - (720,1280,3)
    - Kernel size = (3,3)
    - Max pooling = (2,2)
  - Second layer – 32 filters
  - Third layer – 64 filters
- Flattened – output layer
- Weights of nodes saved
- 10 epochs – one complete presentation of the data to the model



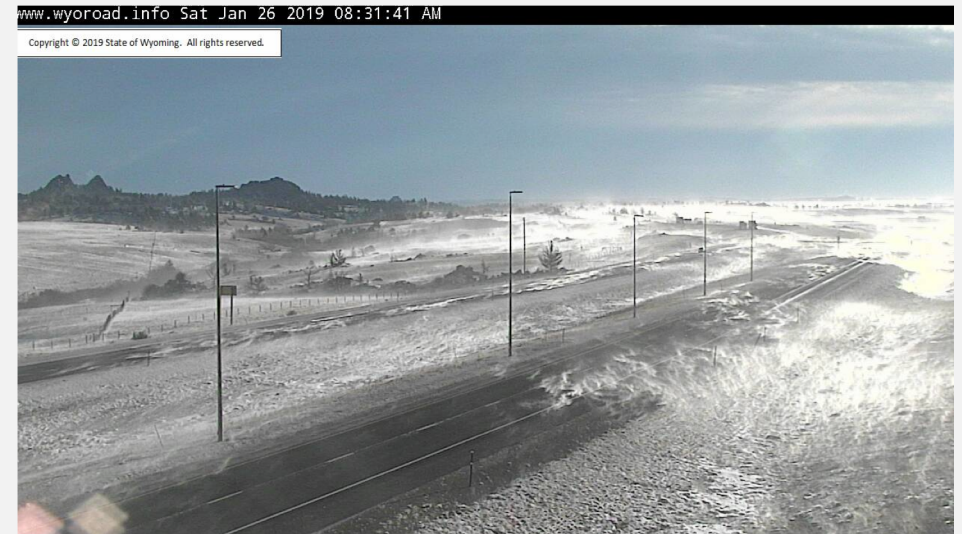
## TRAINING DATASET - BROAD

- Archived WYDOT images
  - October 1, 2018 - January 31, 2019
  - November 1, 2019 - Feb 28, 2020
- Focused along Interstate 80 - 54 cameras
- Angles facing east, west, road sfc

# BINARY MODEL

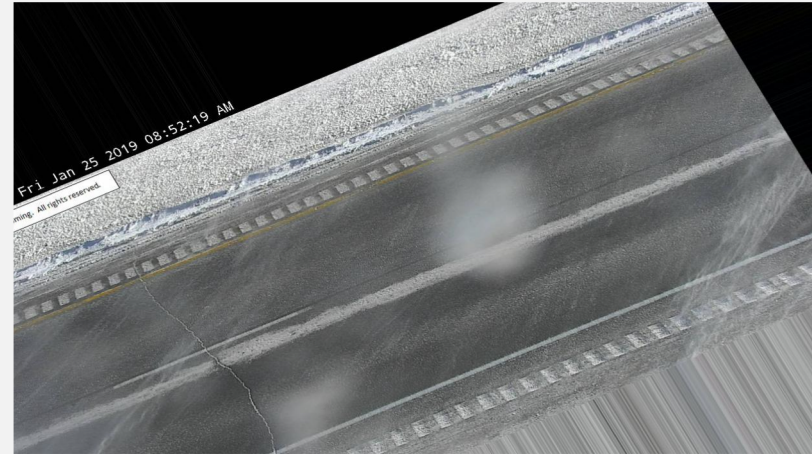
# TRAINING DATASET

- 2 categories – binary model
  - Blowing snow vs not blowing snow (clear road)
  - Multiple camera angles and locations
  - 645 images



# TRAINING DATASET

- Data Augmentation
  - Good for smaller datasets
  - Some road surface images at different angles



## TESTING/VALIDATION DATASET

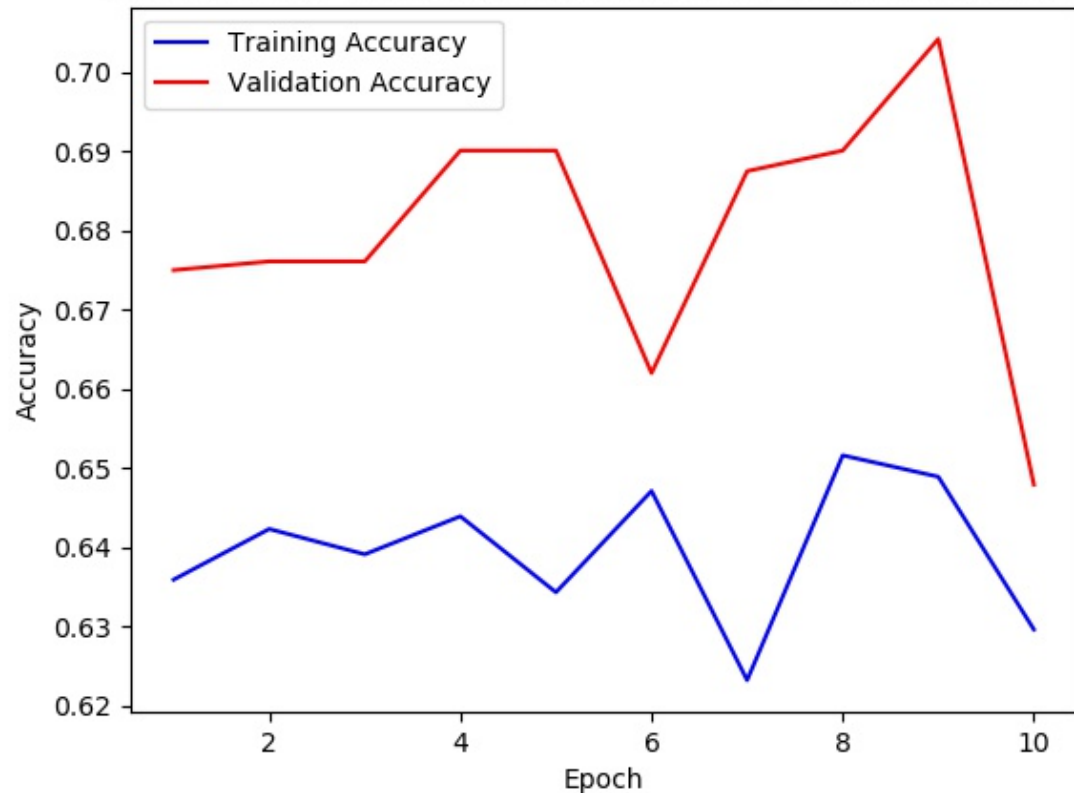
- Provides unbiased evaluation of the model
- Separate images from the training dataset
- 2 categories as well
  - Blowing snow vs not blowing snow
  - 87 images



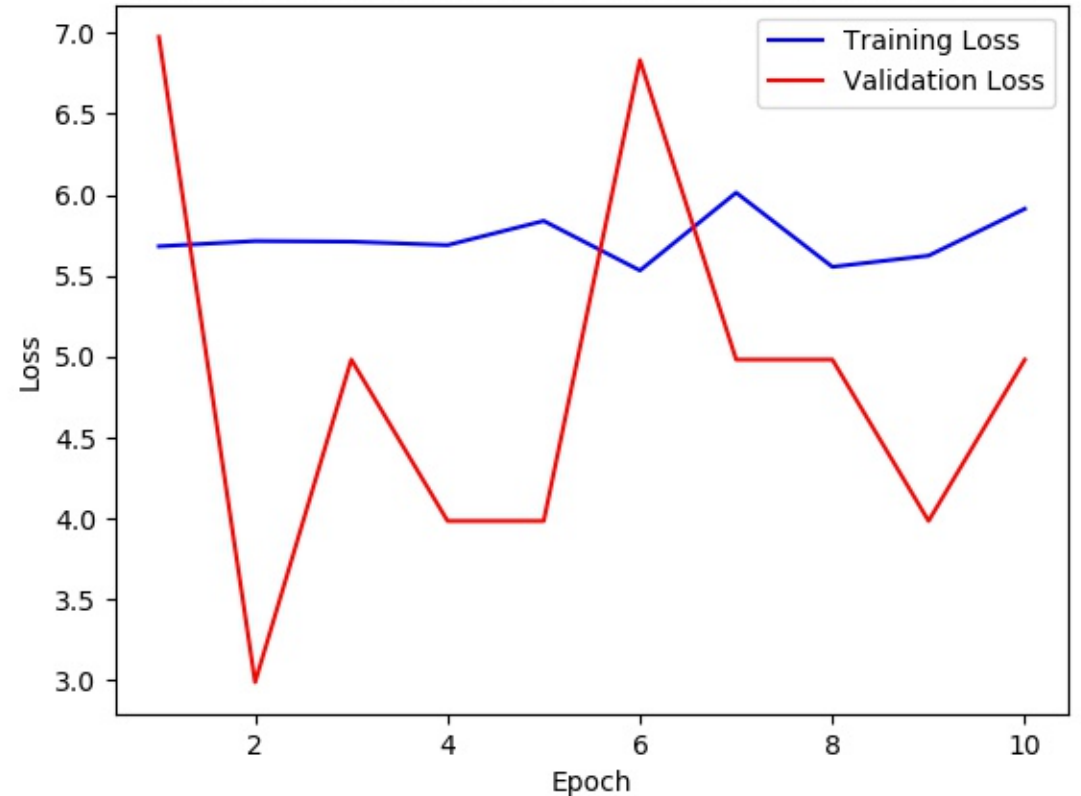
# Binary Model Training: Initial Results

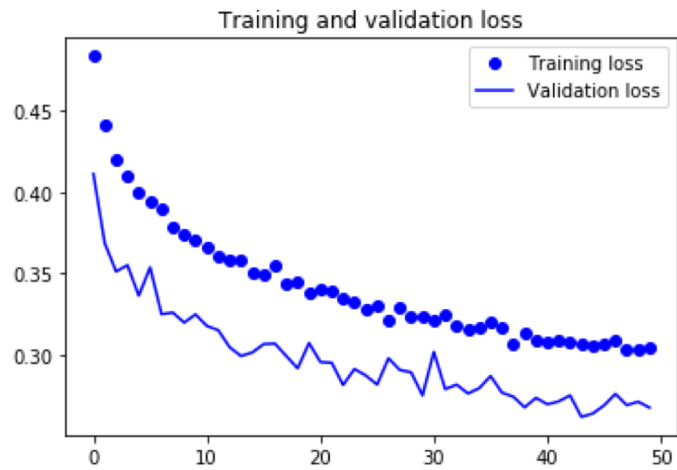
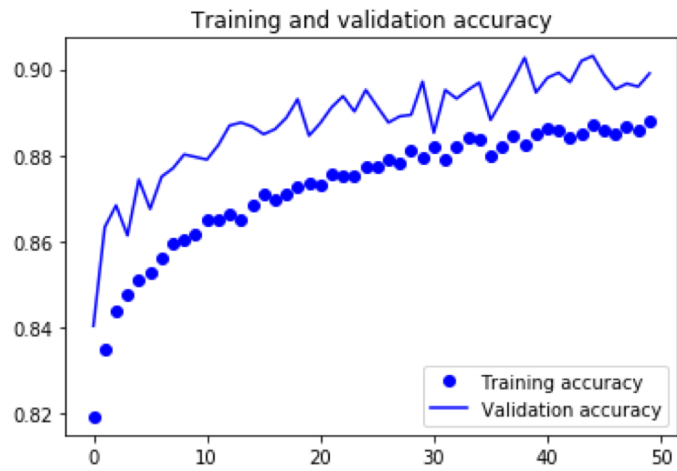
- 645 in training (2 classifications), 87 in validation (2 classifications)
- Accuracies not improving

Blowing Snow Classification: Training and Validation Accuracy



Blowing Snow Classification: Training and Validation Loss

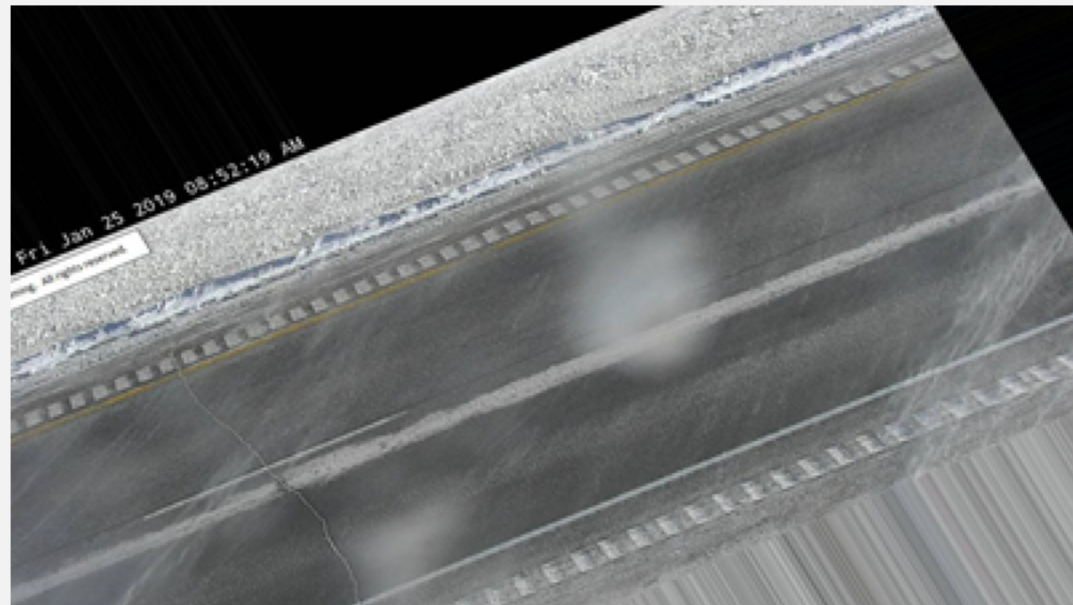
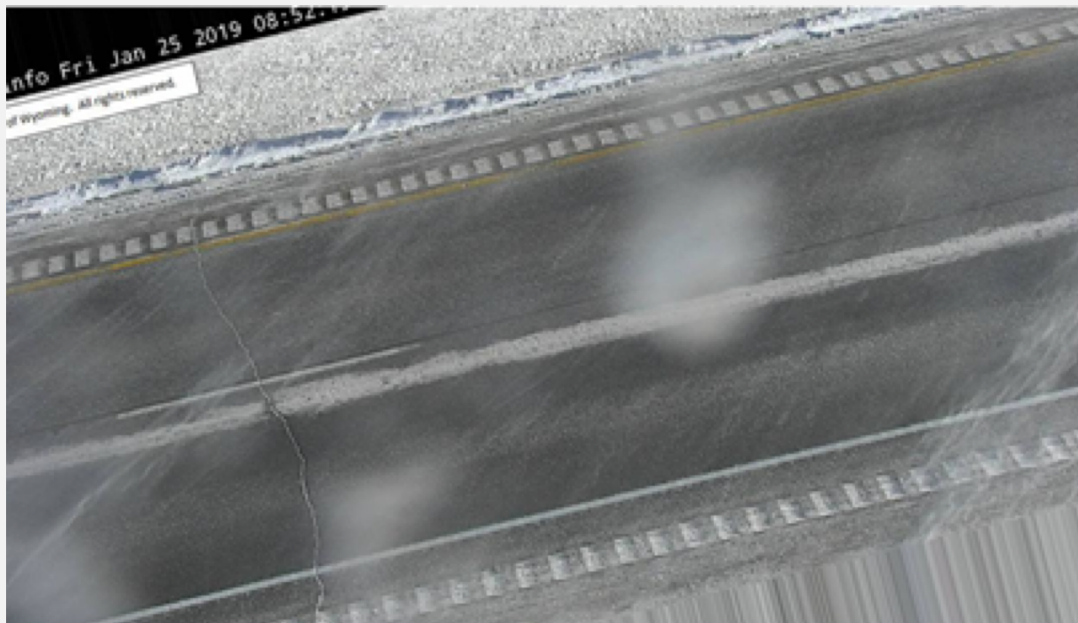




WHAT TO  
EXPECT

# IMPROVING THE TRAINING DATASET

- Got rid rid these images in training dataset:



- Cropped images with cameras facing East and West

# CROP IMAGES

- Example: Vedauwoo East

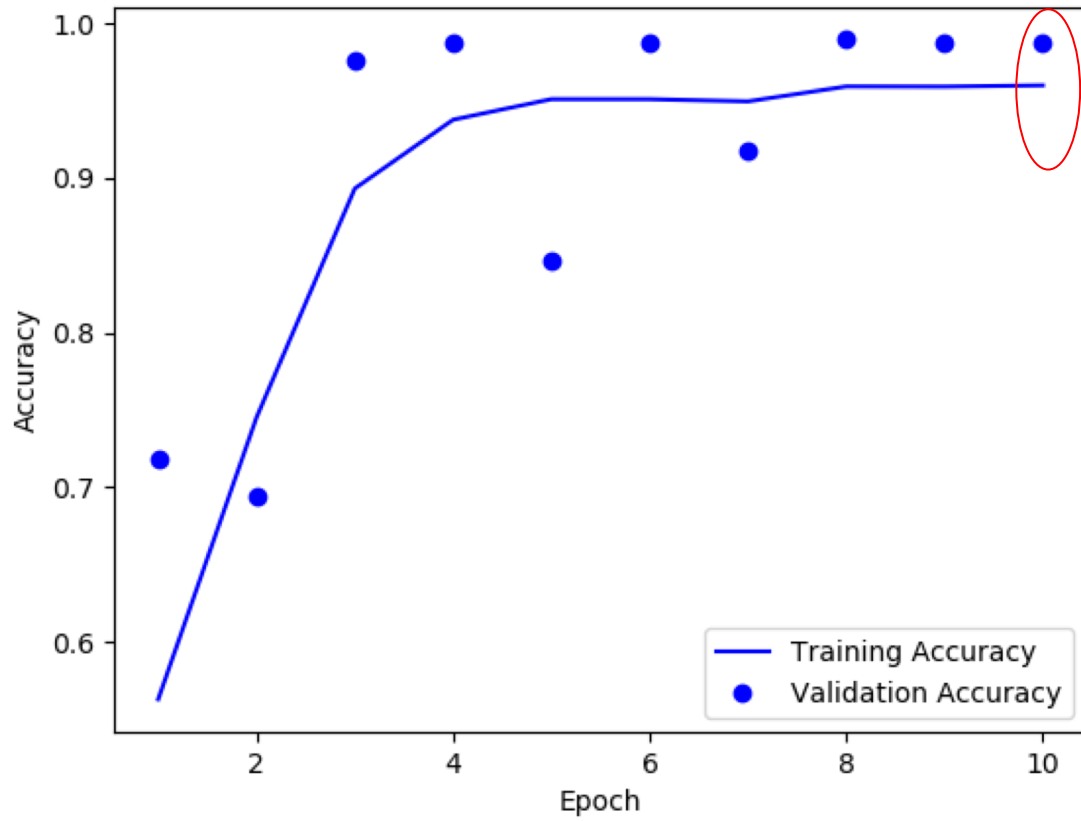


- (720,1280) to (360, 1210)
- Hope that training the model on just the road will improve accuracy

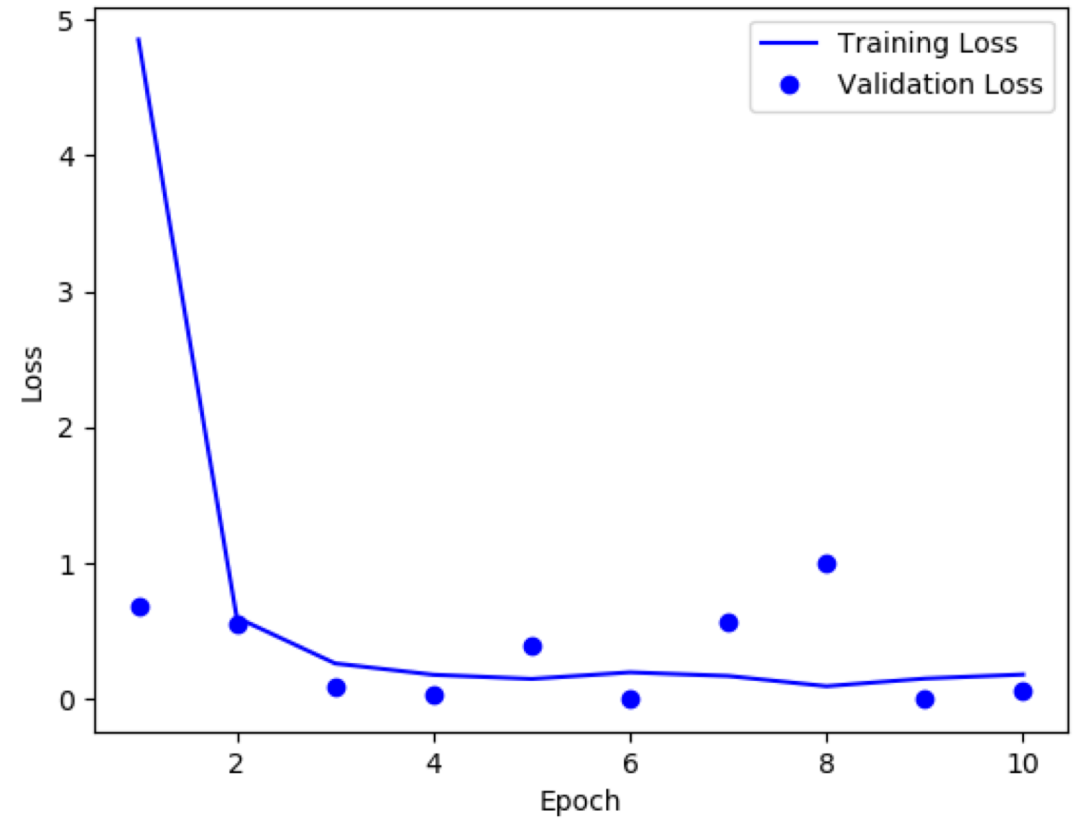
# Improved Accuracy

Training acc = 96.0%  
Validation acc = 98.8%

BSN Classification: Training and Validation Accuracy (Cropped Images)



BSN Classification: Training and Validation Loss (Cropped Images)



# CLASSIFICATION MODEL

# CLASSIFICATION MODEL: BLOWING SNOW INTENSITY CATEGORIES

- Adopted from the Cheyenne NWS office's categories for blowing snow intensity
  - Patchy = Drifting Snow Only
  - Areas = Blowing Snow
  - Def = Advisory Conditions (Near Blizzard) - Sporadic to Occasional whiteout conditions
  - Def 1/4sm = Blizzard Conditions - Persistent whiteout conditions

# CATEGORIAL MODEL TRAINING AND TESTING DATASET

- Training dataset contains 717 images falling into 4 categories:
  - Strong, medium, drifting, no BLSN
- Testing dataset: 178 images
  - Falling into the same categories



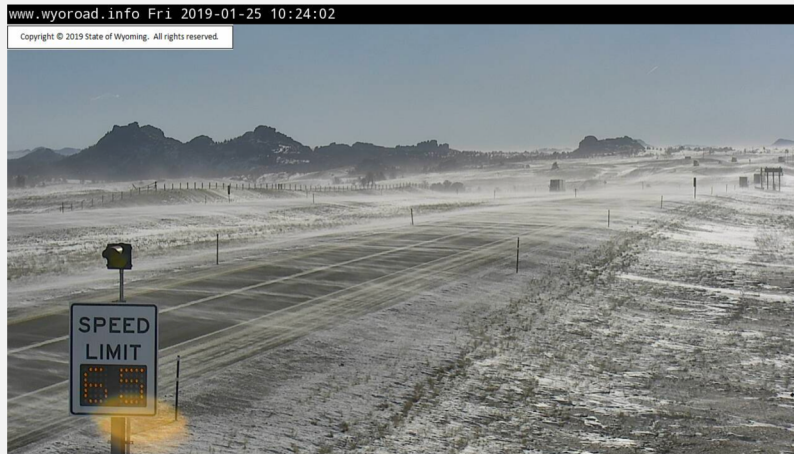
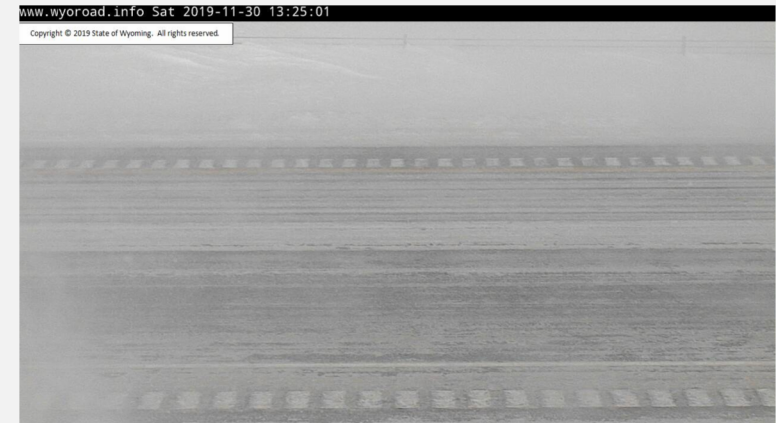
# Drifting



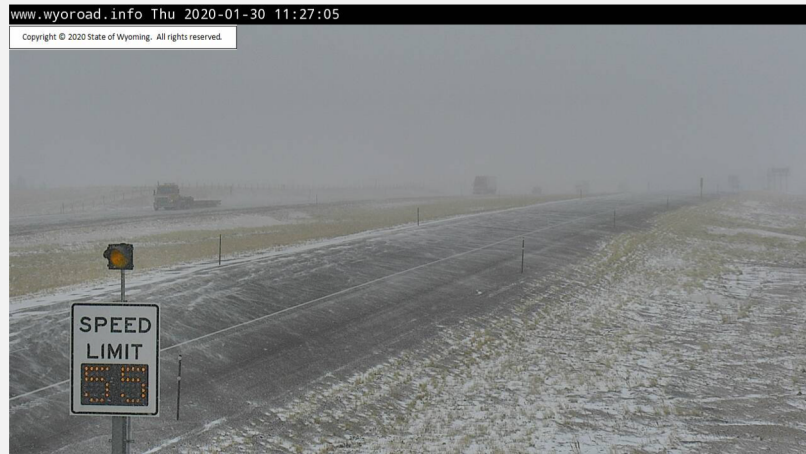
# Medium



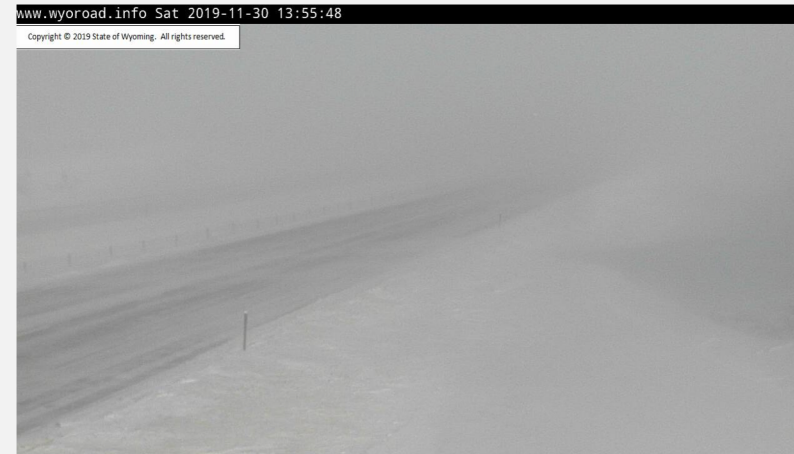
# Strong



Jan 25, 2019



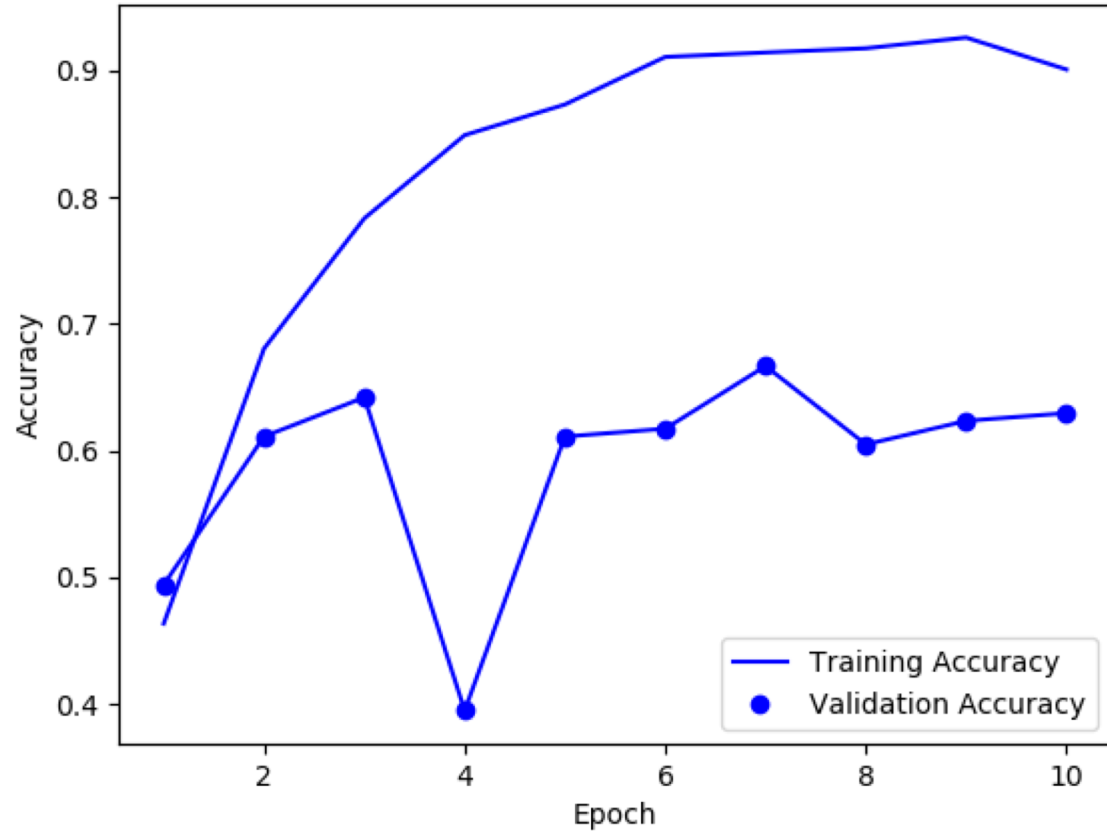
Jan 30, 2020



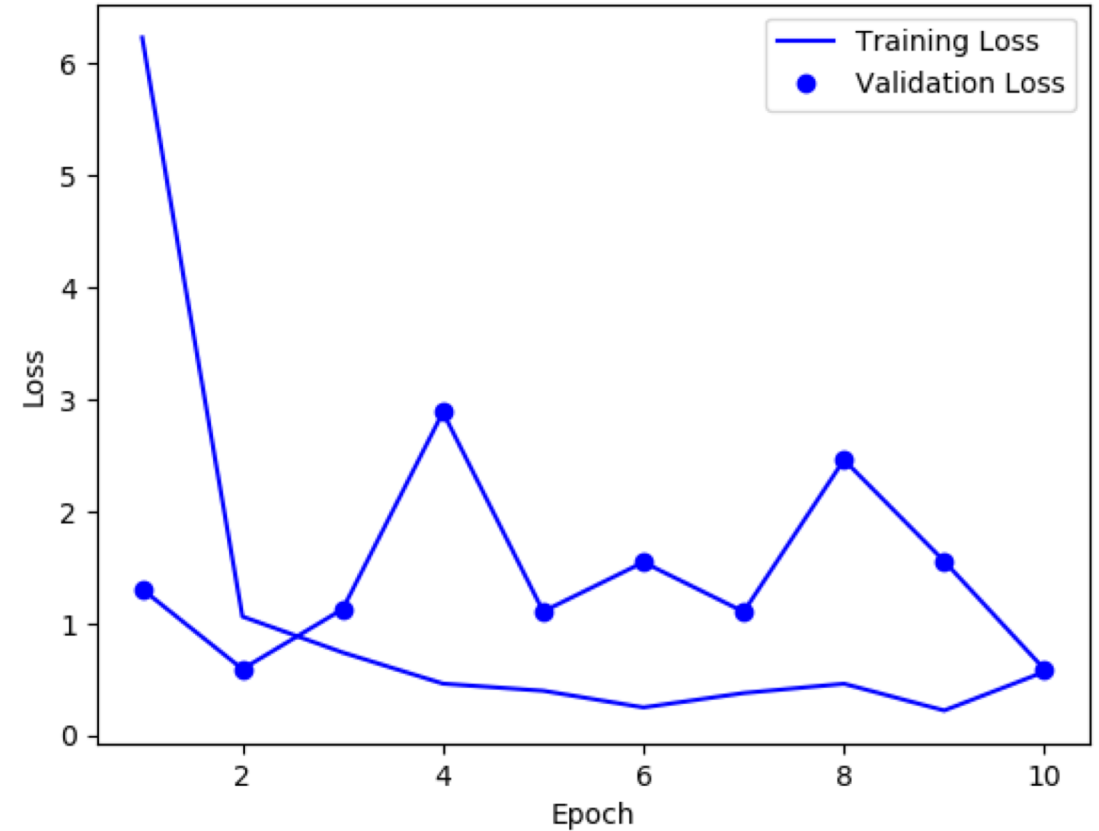
Nov 30, 2019

# Classification Model: Training Results

BSN Intensity Classification: Training and Validation Accuracy



BSN Intensity Classification: Training and Validation Loss



- Training accuracy improves with each epoch
- Validation accuracy not the best → evaluates model performance upon seeing new images
- Training loss drops off nicely, but want it to be close to validation loss
- See possible overfitting here since validation loss > training loss

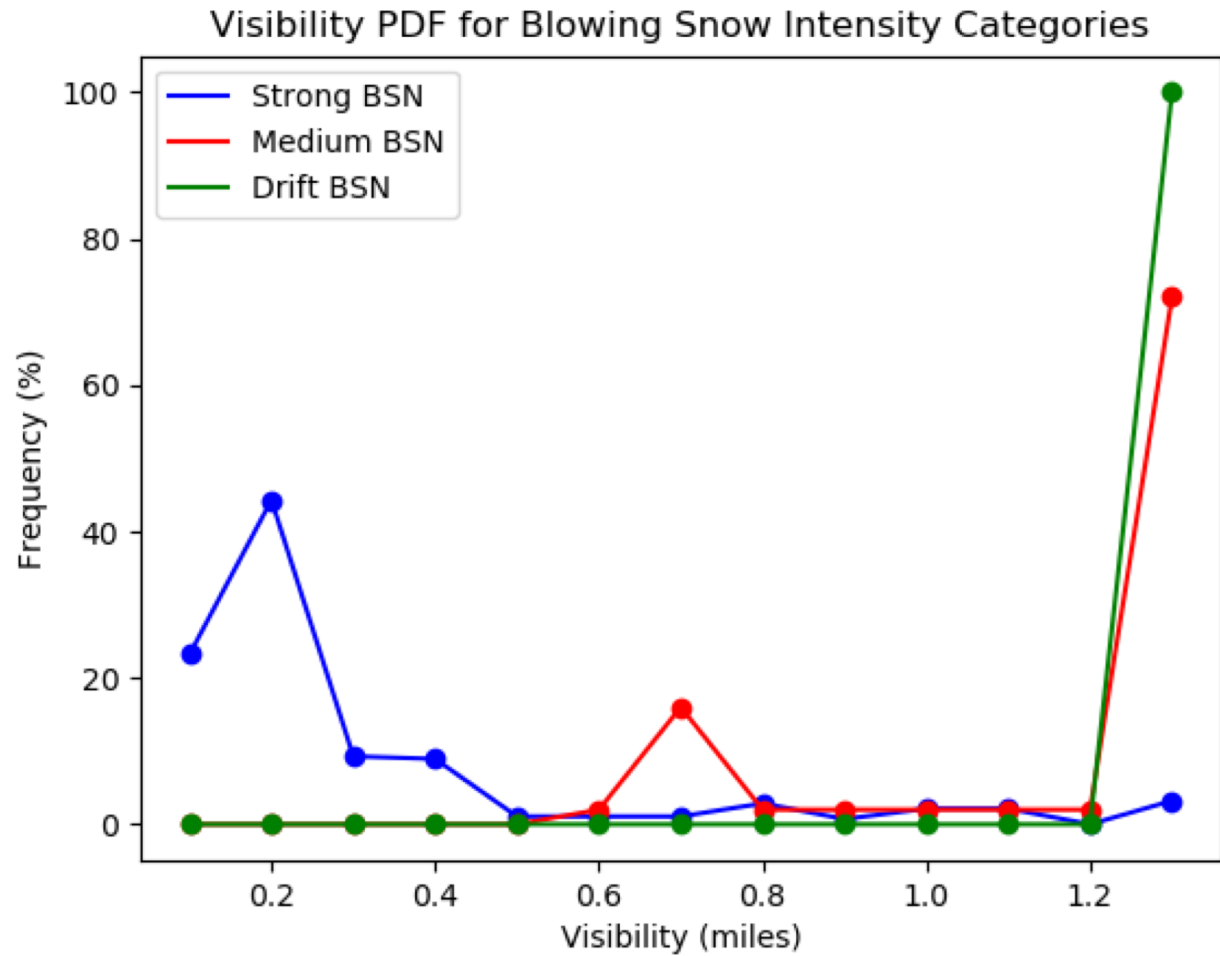
# METEOROLOGICAL CHARACTERISTICS OF EACH CATEGORY

- Gather observations within each category → strong, medium, drifting
  - Air temperature, road temperature, visibility, wind speed, wind gust, RH (liquid), RH (ice)
- Look at probability distributions to see which meteorological variables are relevant for causing BLSN

- Strong: 280 samples
- Medium: 60 samples
- Drift: 80 samples

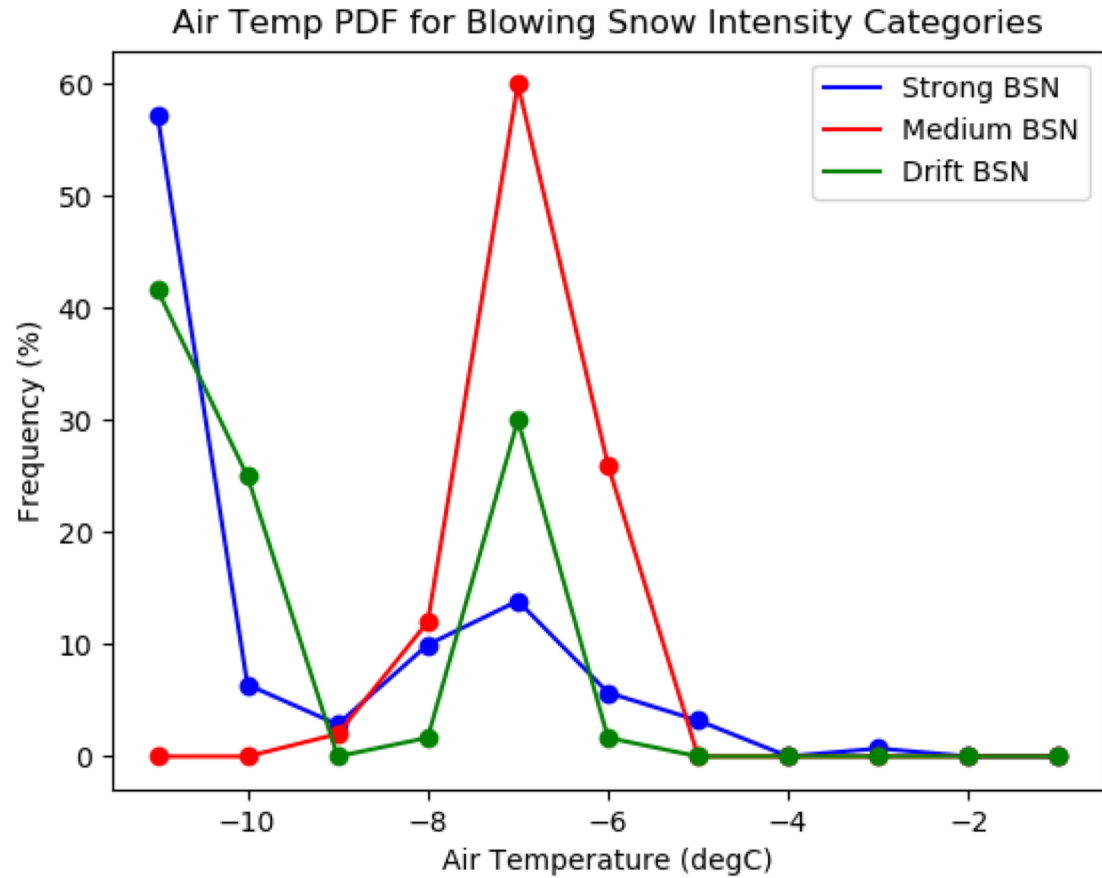


# VISIBILITY PDF

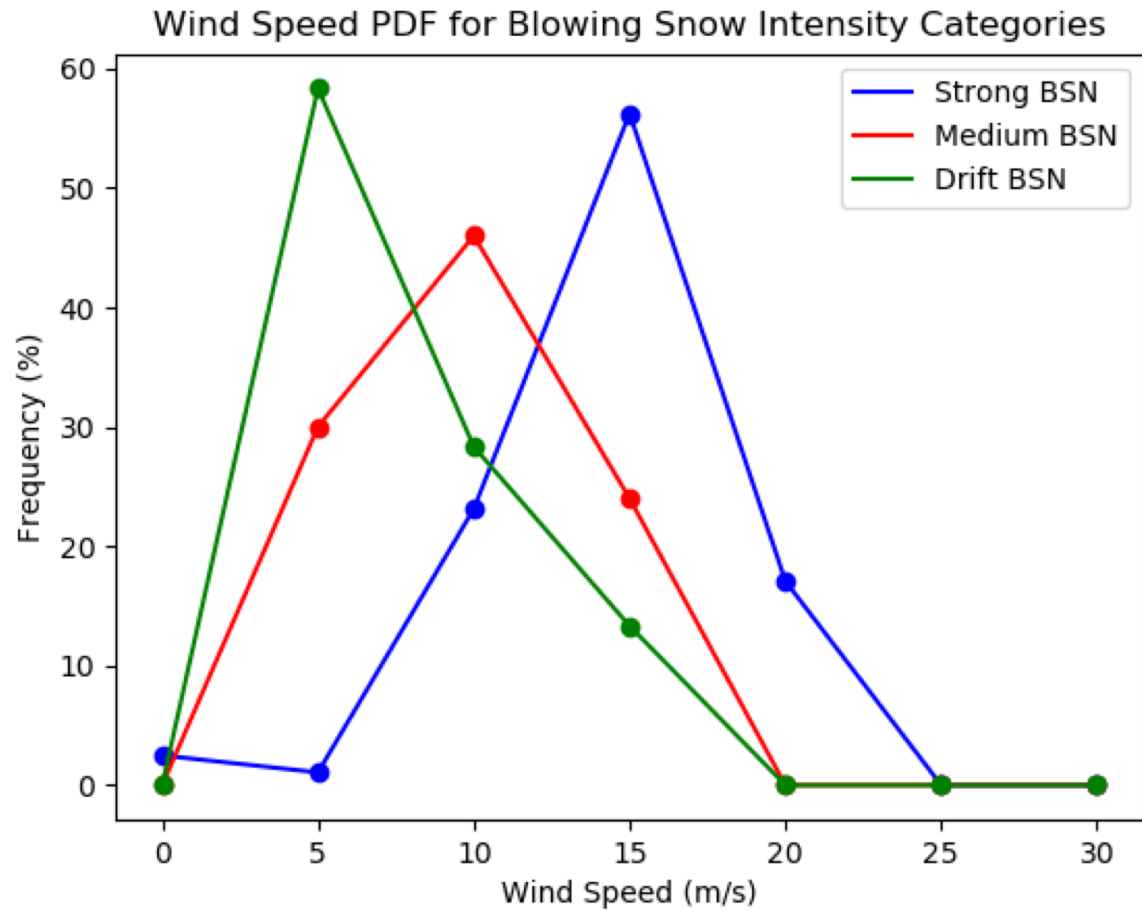


\*1.24 miles is max visibility measured

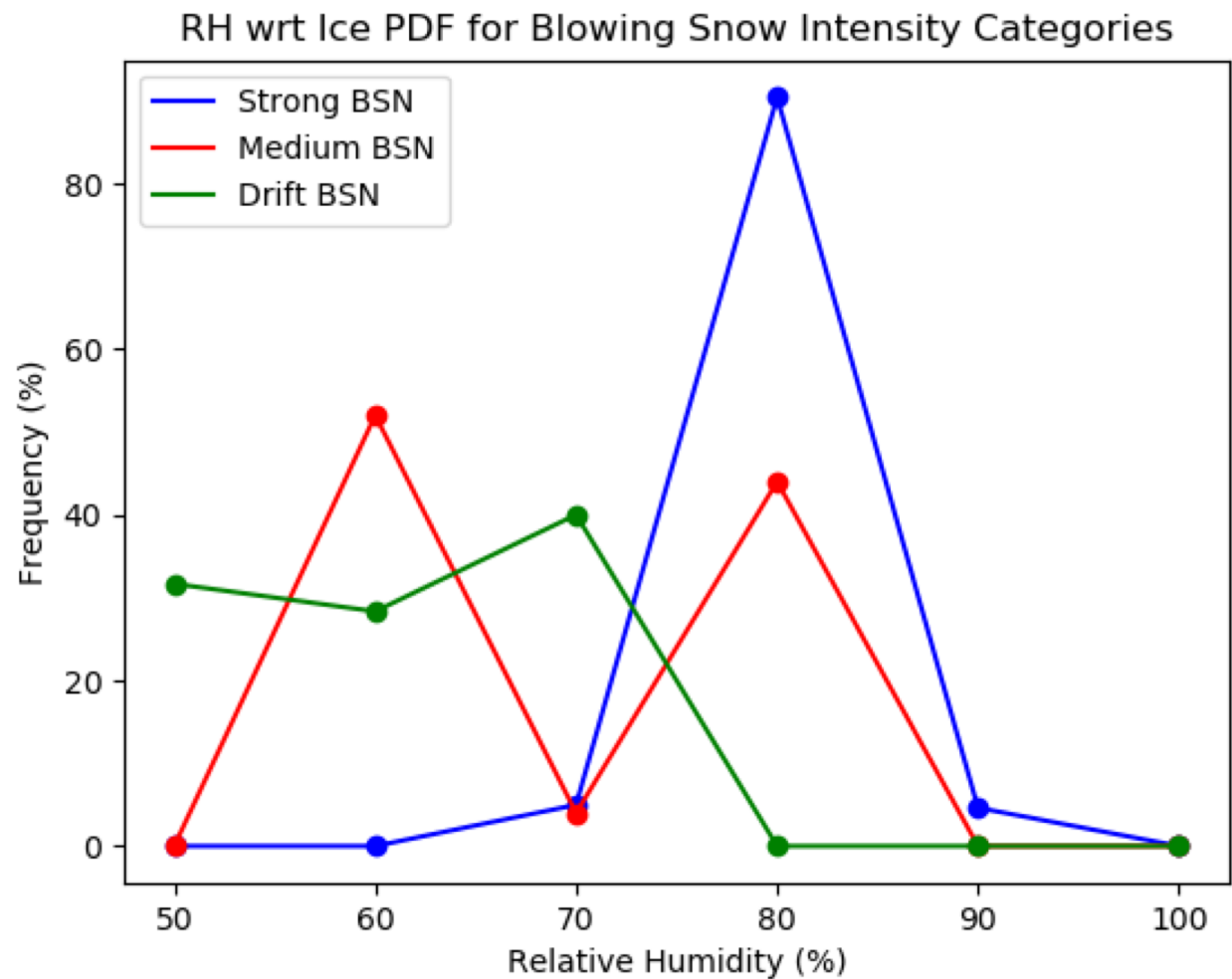
# TEMPERATURE PDF



# WIND PDF

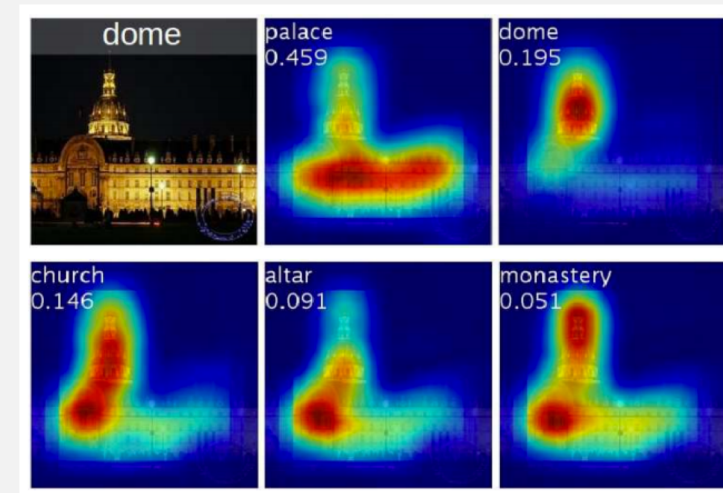


# RELATIVE HUMIDITY PDF



## NEXT STEPS

- Improve model validation accuracy/loss
  - Add more images to dataset
  - Optimize hyperparameters in model
- Create new model with different input types – images and meteorological data to improve accuracies
- Increase model interpretability
  - ML algorithms more transparent, less of a blackbox
  - Use visualizations such as saliency maps, class-activation maps, etc.





## Thank you to:

- Zach Lebo
- CSTAR lab group
- WYDOT for the archived images
- Cheyenne NWS
- NOAA Grant NA19NWS4680005

# Questions?

Q&A session Monday, July 13<sup>th</sup> at 3:15 MT

Contact:  
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