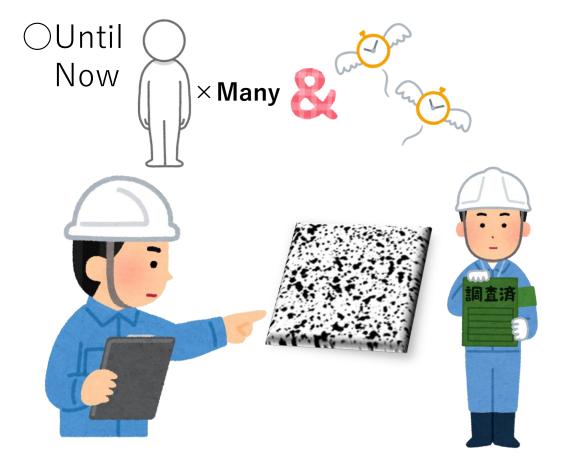
A Study on the Evaluation of Rust on Weathering Steel Using Machine Learning

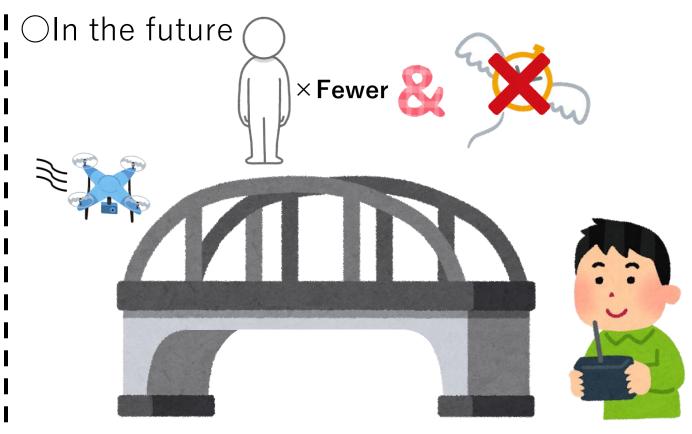
7022615 Ren Morii

 If we can build a system that can look at the rust and give consistent scores, inspections will become more reliable and efficient.

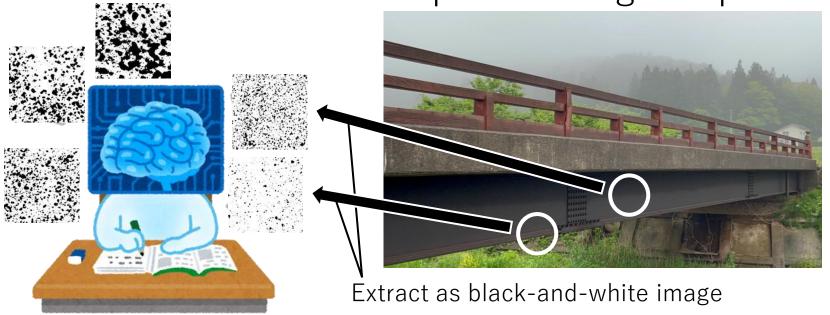


 If we can build a system that can look at the rust and give consistent scores, inspections will become more reliable and efficient.

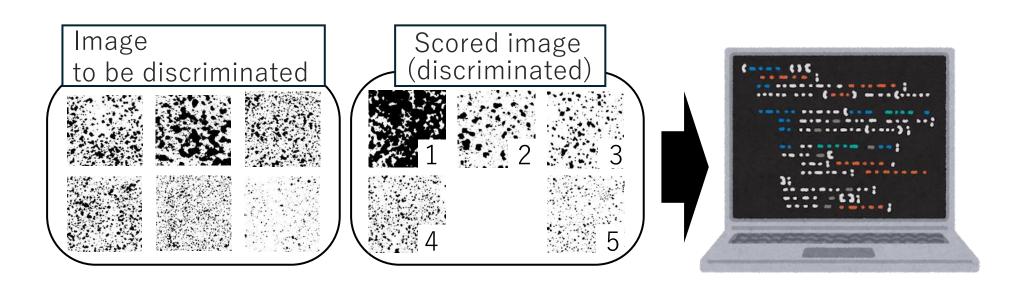




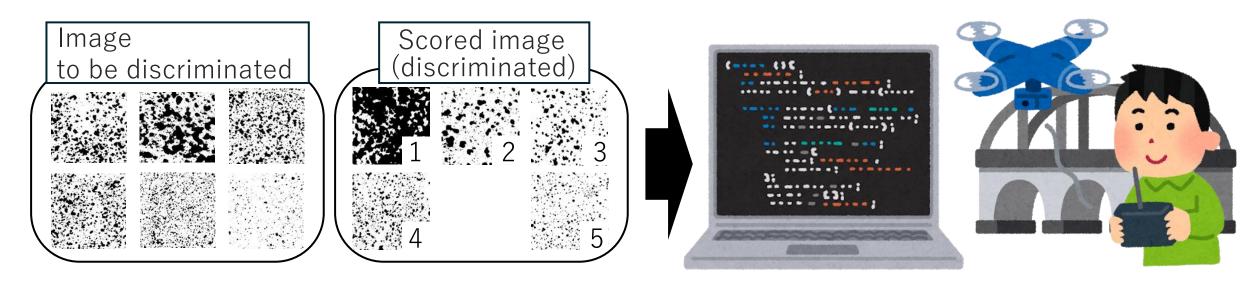
- My graduation research is …
 To automatically evaluate the condition of rust on weathering steel bridges.
- The goal of this research is ···
 To create an Al system that can analyze images taken by drones to help with bridge inspections.



- I am working with rust images collected from past field investigations and developing a machine learning model that can look at these images and assign them a rust severity score, from 1(very rusty = bad) to 5(very clean = good).
- This helps make infrastructure inspections safer and more consistent.



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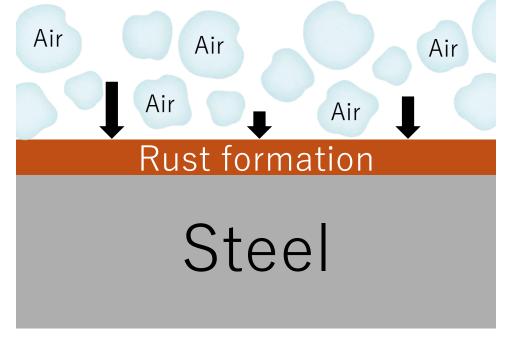


- Weathering Steel is a type of steel designed to reduce the need for painting and periodic maintenance.
- By pre-dressing the surface with rust, it acts as a protective layer that protects the interior and prevents corrosion.



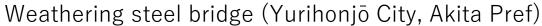


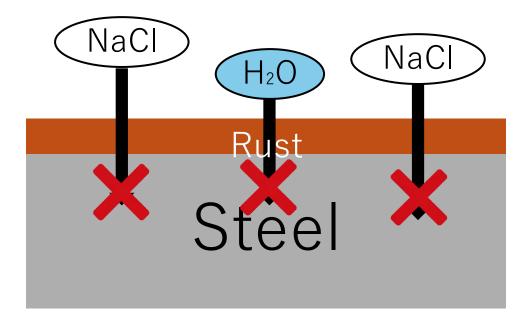
Weathering steel bridge (Yurihonjō City, Akita Pref)



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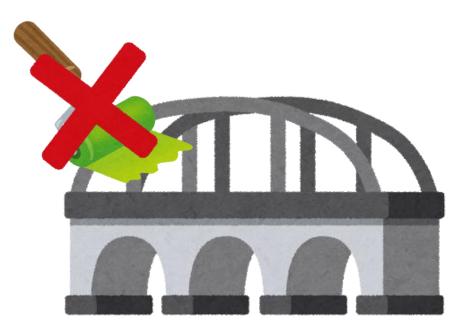






○ Merit

- Reduced maintenance costs.
- ⇒It does not need to be painted, except in areas facing the sea, thus reducing the time and cost required for painting.
- It is used for bridges and buildings not only in Japan but also in other countries.



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Pinqui - Work created by the contributor, CC Attribution-ShareAlike 3.0, https://commons.wikimedia.org/w/index.php?curid=12224155による

Student dormitory at a university in the UK

Tsunami Disaster Prevention Station on Awaji Island, Hyōgo

Pref



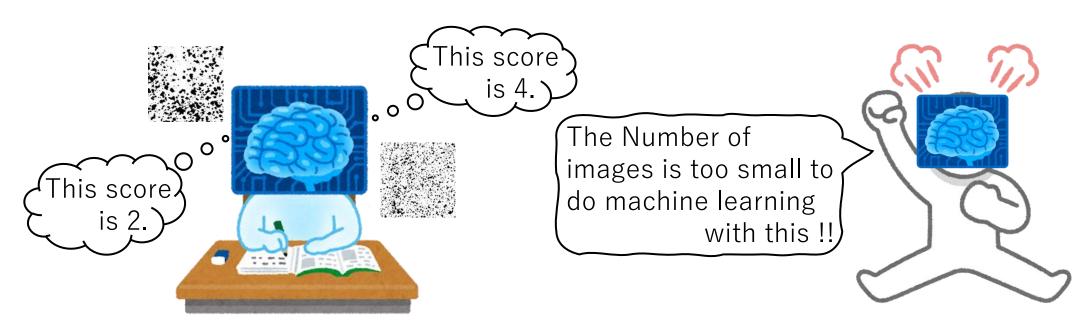
ex.php?curid=8113650による

OThings to watch out for

- The rust protection layer works only when the surface condition is properly managed, and if the rust grows too much or peels off, the service life will be reduced.
- Periodic inspections and surface coating are necessary in humid areas or areas where salt is blown in,



- In this project
 I use AI to give rust images a score between 1 (bad) and 5 (good).
- I start with a small number of images that have already been scored by senior. There are 24 labeled rust images. These labels tell the AI what the "correct" score is.

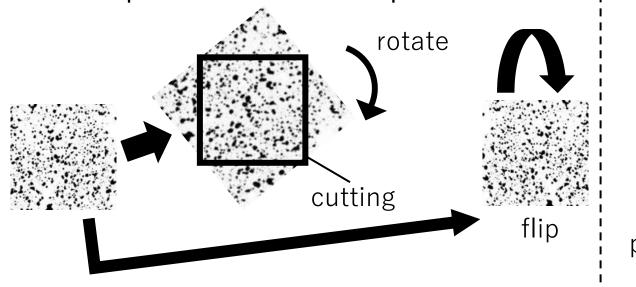


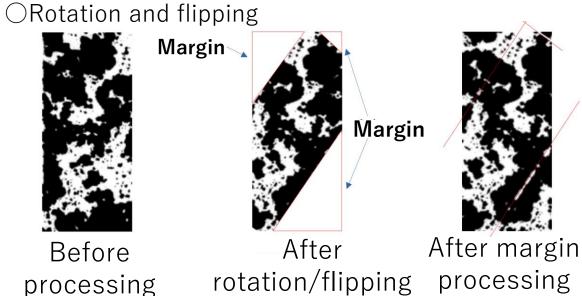
 To solve this problem using data augmentation. This means we create many new images from the original 24.

ex) rotate the image, flip it, or combine rotation and flipping

⇒ The number of images could be increased from 24 labeled rust images to approximately 4,000 by using data

expansion techniques.





- · CNNs are good at recognizing features in images. I tried different models, such as EfficientNetV2, DenseNet121, and MobileNetV3. These models were first trained on general images, and I retrain them with our rust images.
- The CNN looks at a rust image and outputs a score, like 3.45 or 4.10. Later, I round this score into one of five categories using a function called binning.

Rust image (example)

Rust image (example)

Rust sampling, scanning & learning

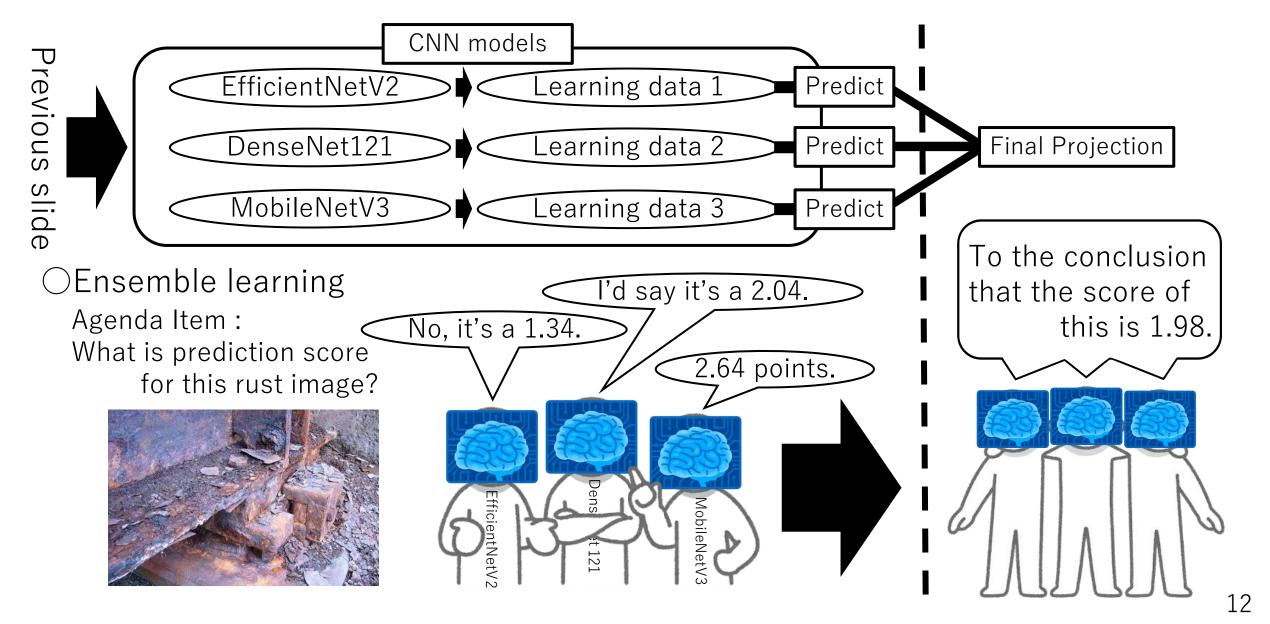
MobileNetV3

Retrieved from https://www.nilim.go.jp/lab/bcg/siryou/tnn/tnn0828pdf/ks082805.pdf

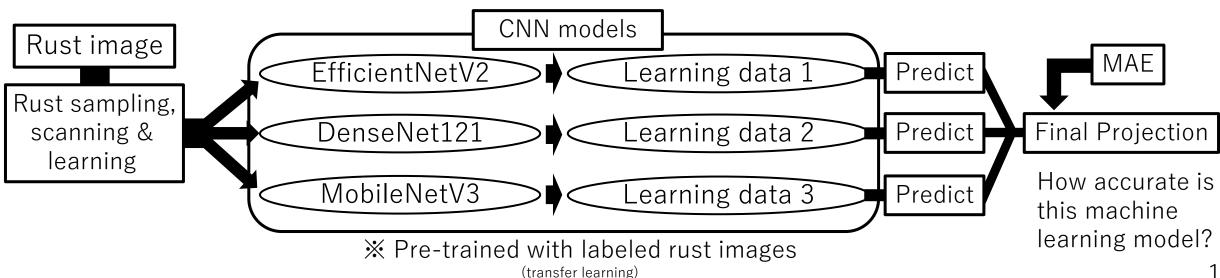
Rust sampling, scanning & learning

MobileNetV3

**Pre-trained with labeled rust images (transfer learning)



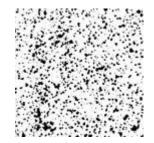
1, Training multiple models (=EfficientNetV2, DenseNet121, MobileNetV3) Each model learns in a slightly different way. After training, I check how well each model can guess the rust score. I use a measure called MAE — Mean Absolute Error — to check the accuracy.



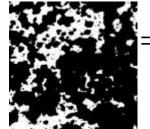
13

- 2, Using visual features
 - Rust area ratio: What percent of the image is rust?
 - Rust count : How many rust spots are there?
 - Max rust size : How big is the largest rust patch?
 - ⇒ I make small adjustments to the Al's score.

Example



⇒Lots of rust



⇒High rust area raito



Max area of rust is large

and tally the number of its rings. Divide by total pixels.

Extract the contour Rust areas (black) are extracted by pixels.

2, In python···

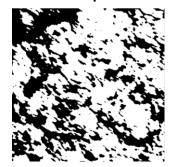
```
# Score correction function by features
def adjust_score_by_features(score, rust_ratio, rust_count, max_blob_area):
   correction = 0.0
                                                     # Maximum rust size
   # rust ratio
                                                     if max_blob_area \geq 7500:
   if rust ratio >= 0.50:
                                                           correction -= 0.5
      correction -= 0.5
                                                     elif 2500 <= rust_ratio < 7500:
   elif 0.25 <= rust_ratio < 0.50:
      correction -= 0.2
                                                           correction -= 0.25
                                                     elif rust_ratio < 2500:
   # Number of rust : 2632 (Pixel)
                                                           correction += 0.1
   if rust_count >= 800:
      correction += 0.3
                                                     return score + correction
```

- Binning
 - \bigcirc If the predicted score is less than 1.9 \Rightarrow Rating 1
 - \bigcirc If the predicted score is between 1.9 and 2.5 \Rightarrow Rating 2
 - \bigcirc If the predicted score is between 2.5 and 3.1 \Rightarrow Rating 3
 - \bigcirc If the predicted score is between 3.1 and 3.7 \Rightarrow Rating 4
 - \bigcirc If the predicted score is greater than 3.7 \Rightarrow Rating 5

2,

```
# Score correction function by features
def adjust_score_by_features(score, rust_ratio, rust_count, max_blob_area):
   correction = 0.0
                                                     # Maximum rust size
   # rust ratio
                                                     if max\_blob\_area >= 7500:
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                                                     elif rust_ratio < 2500:
                                                           correction += 0.1
   if rust_count >= 800:
                                                     return score + correction
      correction += 0.3
```

Example



Regression Score: 2.71 (Rating 3)

Max rust size : 2632 (Pixel)



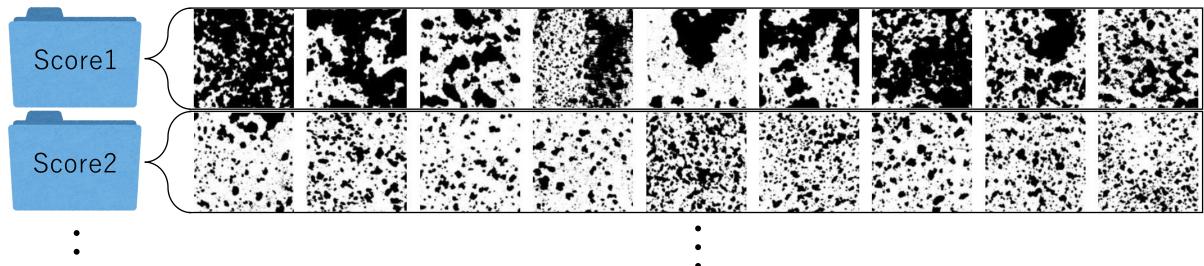
Corrected Score: 2.46 (Rating 2)



The Max rust size is relatively large.

Let's subtract 0.25 points from your score.

- 3, Organizing the images into folders
 - This makes it easy to distinguish between images of light and heavy rust.
 - Rust comparisons can be made for each of the same scores to determine that this is a rust image that should be stored in a different score, leading to better programming.

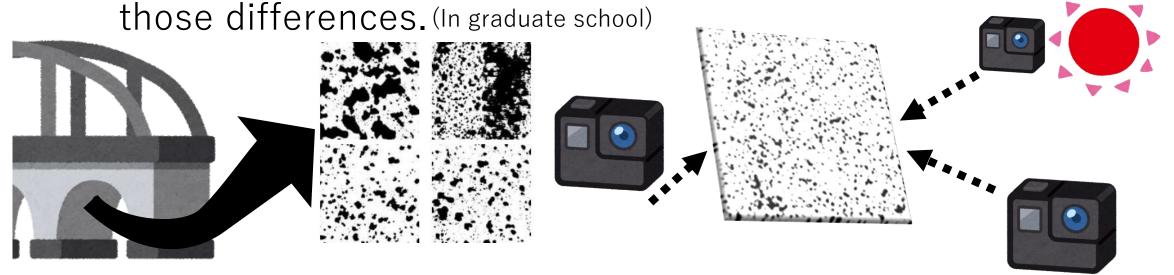


- 3, Writing logs that show how the AI makes decisions, including the original score, adjusted score, and final score. This helps me understand why a certain score was given.
- logs

⇒ Rewrite the feature functions shown in the previous to build a better discriminant model, and then display MAE to show which CNN models can be combined for better accuracy.

5, Challenges and Future Goals

- · Challenge1: The small number of labeled images
- ⇒ More-labeled data will help the Al learn better and I hope to join future field investigations and collect more images myself.
- Challenge2: To use drone images
- ⇒ These images will be taken from different heights and angles, so the Al model must be strong enough to handle



5, Challenges and Future Goals

- Creating a simple app or web system
 - ⇒This allows engineers to load rusty images and instantly get an automatic score.
- My dream is to use technology to support infrastructure inspection, reduce human error, and protect public safety.

References

- [1] 国土技術政策総合研究所, 耐候性鋼橋の表面性状の事例とさびの外観性状, https://www.nilim.go.jp/lab/bcg/siryou/tnn/tnn0828pdf/ks082805.pdf (Viewed July 18, 2025)
- [2] 耐候性鋼, (Wikipedia, the free encyclopedia) http://ja.Wikipedia.org/wiki/耐候性鋼. (Viewed July 28, 2025, from the most recent version)