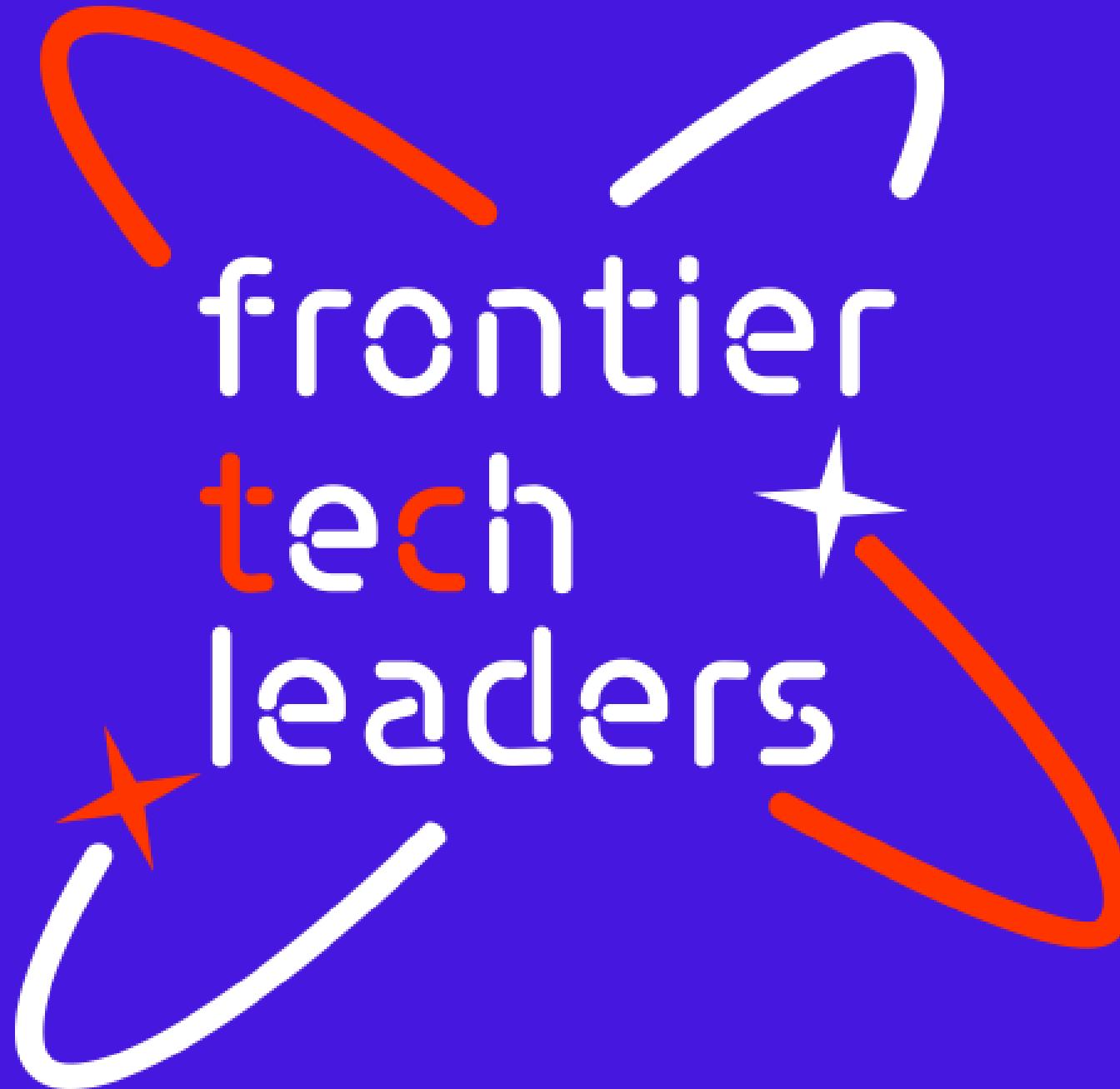




A new generation
of tech **specialists**



Snake Species Identification

Team Members :
Pyae Sone Kyaw
Htwe Myat Cho
Aye Moh Moh Khin
San San Maw

Background

- Snakebite incidents remain a major health risk, especially in rural areas.
- Delayed or incorrect snake identification leads to improper treatment.
- Existing tools lack intelligent, real-time species recognition.
- AI/ML-powered identification can reduce preventable injury and improve emergency response.

Problem Importance

- Non-experts struggle to differentiate venomous vs. non-venomous species.
- Misinformation during snake encounters increases panic and risk.
- Limited access to snake experts or reliable field guides.
- Automated identification + safety guidance can save lives.

Objectives

- Develop a CNN model to classify snake species from images.
- Automatically determine venomous status.
- Generate a real-time, species-specific safety card.
- Improve public awareness and speed of emergency decision-making.

SDG Relation

SDG 3 – Good Health & Well-being

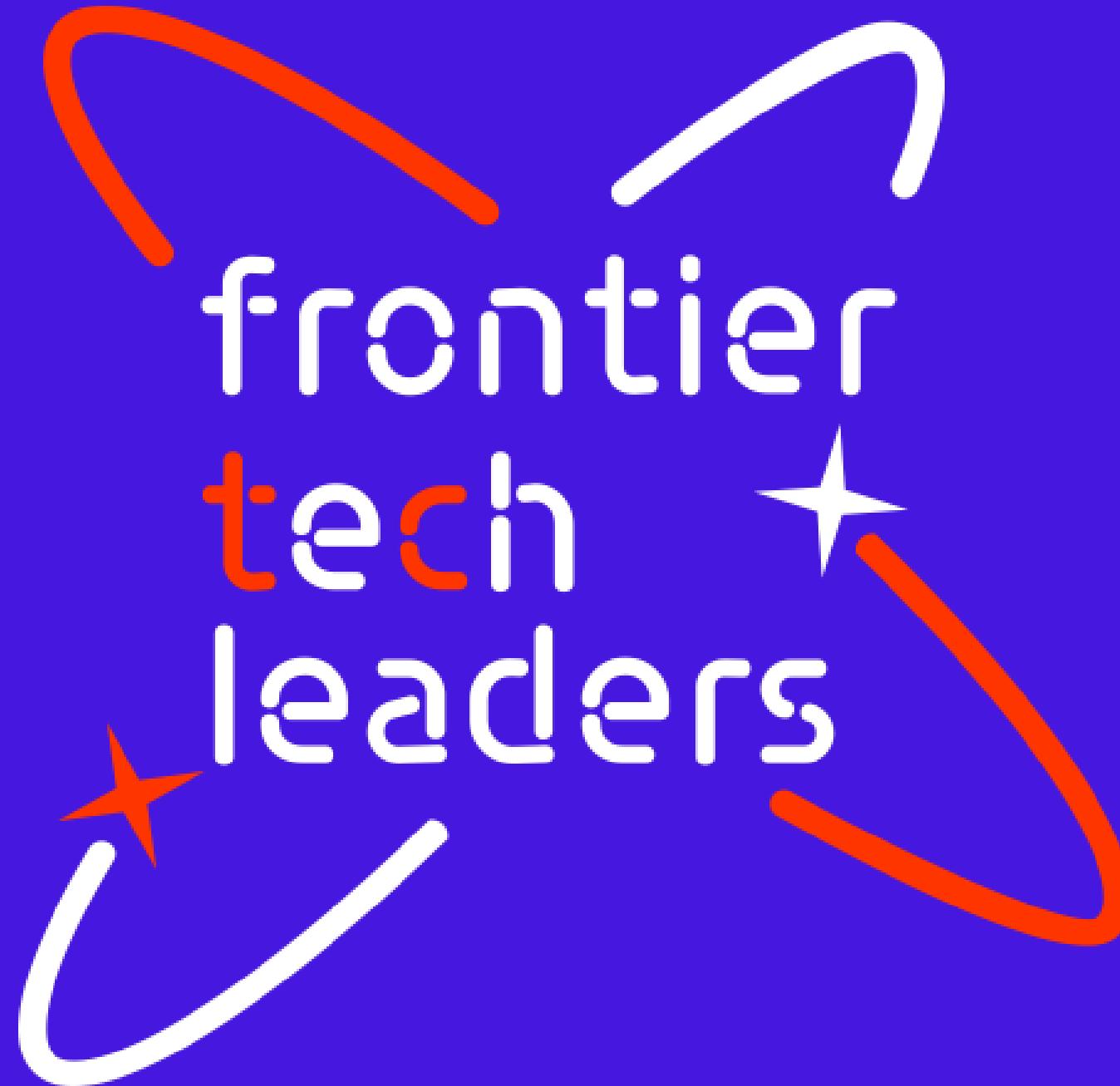
- Reduce preventable snakebite morbidity through instant guidance.

SDG 9 – Industry, Innovation & Infrastructure

- Apply deep learning for public safety tools and digital resilience.

SDG 11 – Sustainable Cities & Communities

- Promote safer human–wildlife co-existence and community preparedness.



Data



Data Sources

- SnakeCLEF 2021: 300k+ images, 700+ species globally.
- iNaturalist: Regional snake observations with verified labels.
- HerpMapper: Supplementary local species images.
- Open licensing ensured (CC-BY / CC0).

Data Collection & Cleaning

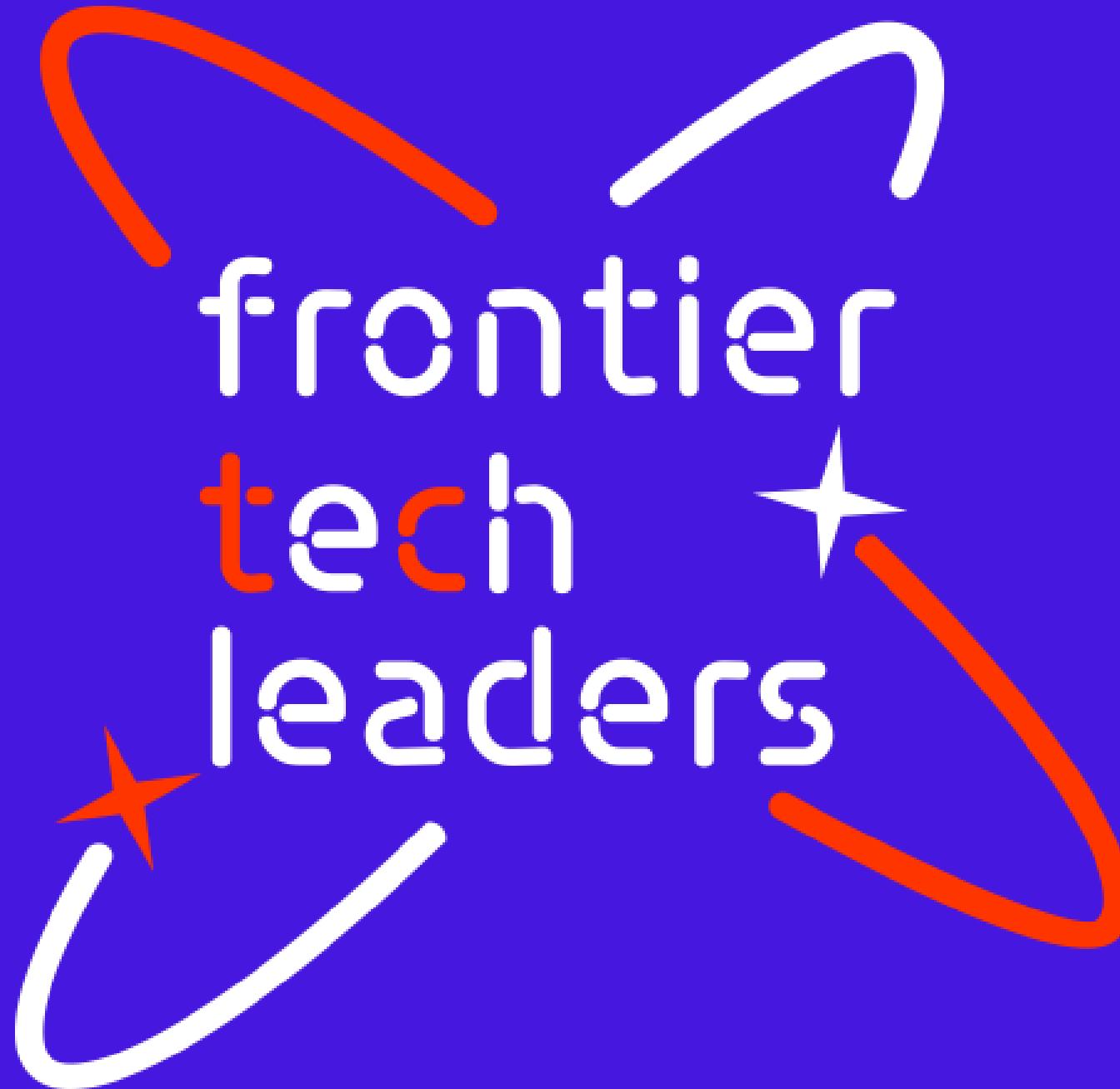
- Created species list (`species_list.csv`) to define targets.
- Scripted pipeline: download images, remove duplicates, ensure valid licensing.
- Offline cleaning: `PIL.Image.verify()`, resize to 256x256, RGB format.
- Runtime cleaning: `SafeImageFolder` skips corrupted files.

Exploratory Data Analysis (EDA)

- Dataset structure: number of species, images per class.
- Image quality: resolution, lighting, background clutter.
- Detected class imbalance: common vs rare species.
- Outcome: balanced, clean dataset ready for CNN training.

Feature Engineering & Transformation

- Model-based: EfficientNet / MobileNet automatically extract hierarchical features.
- Color augmentation (brightness, contrast, hue).
- Spatial augmentation (flip, crop, rotation).
- Normalize using ImageNet mean/std.
- Convert images to tensors for CNN input.



Model

Model Selection

- MobileNetV3-Small: lightweight, fast, good for edge deployment.
- EfficientNet-B0: higher accuracy, balanced compute requirements.
- Strengths: small size, fast inference, easy ONNX export.
- Weaknesses: may underfit fine-grained species cues; sensitive to resolution.

Model Training

Two-stage transfer learning:

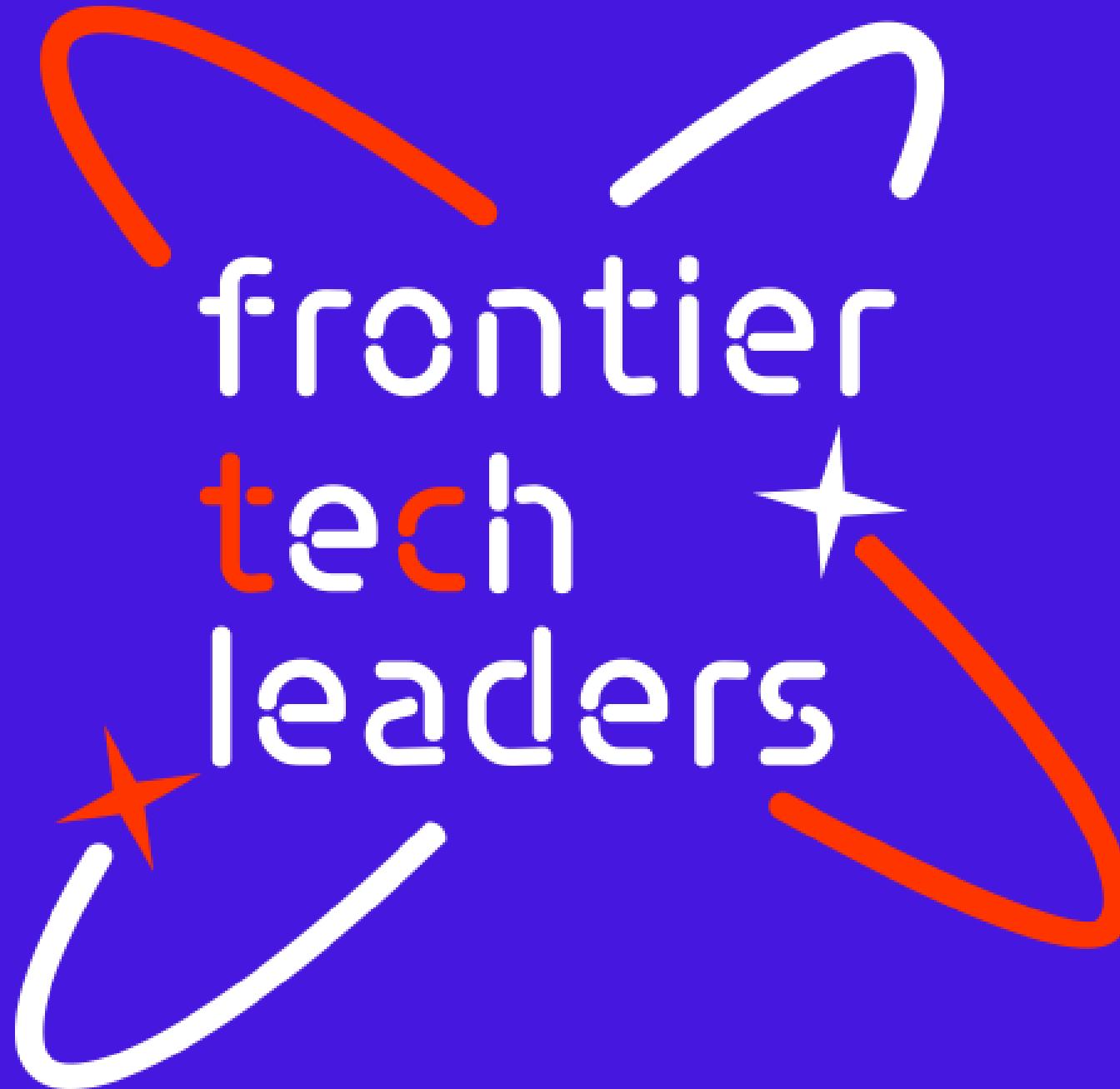
- Head Training: train only classification head, backbone frozen.
 - LR: 1e-3, AdamW, Cross-Entropy, 10 epochs.
- Fine-Tuning: unfreeze last backbone block.
 - LR: 5e-4, 5 epochs minimum 2.

Data: processed images, augmented, train/val split 80/20.

Batch size: 64

Model Refinement

- Stronger augmentations, SafeImageFolder for robustness.
- Partial fine-tuning of EfficientNet-B0 / MobileNetV3-Small.
- Learning rate schedule: head $1e-3 \rightarrow$ fine-tune $5e-4$.
- Batch size tuned (32–64), no full k-fold CV due to time constraints.



Result

Model Evaluation

Test Accuracy: 0.5667

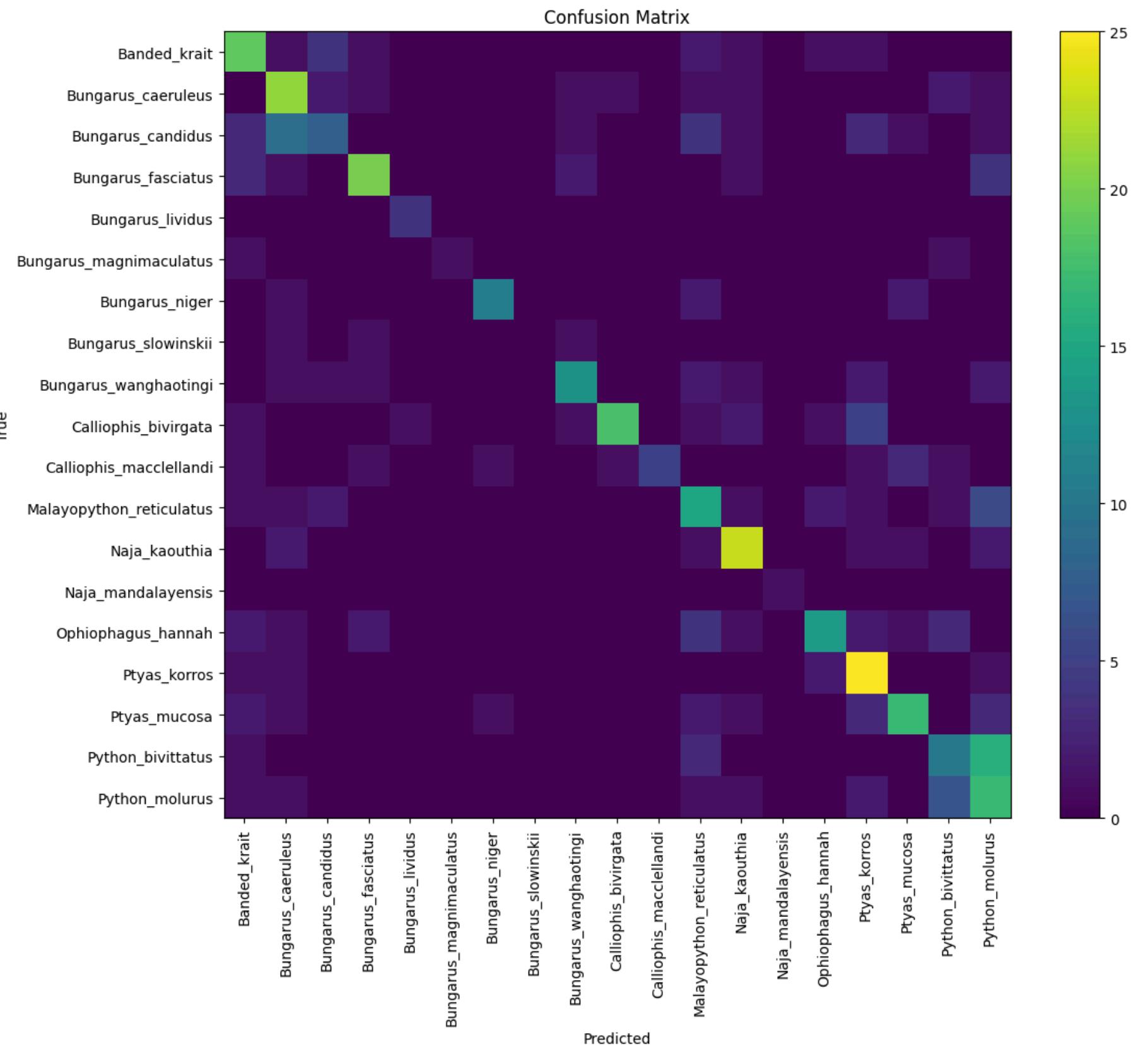
Classification Report:

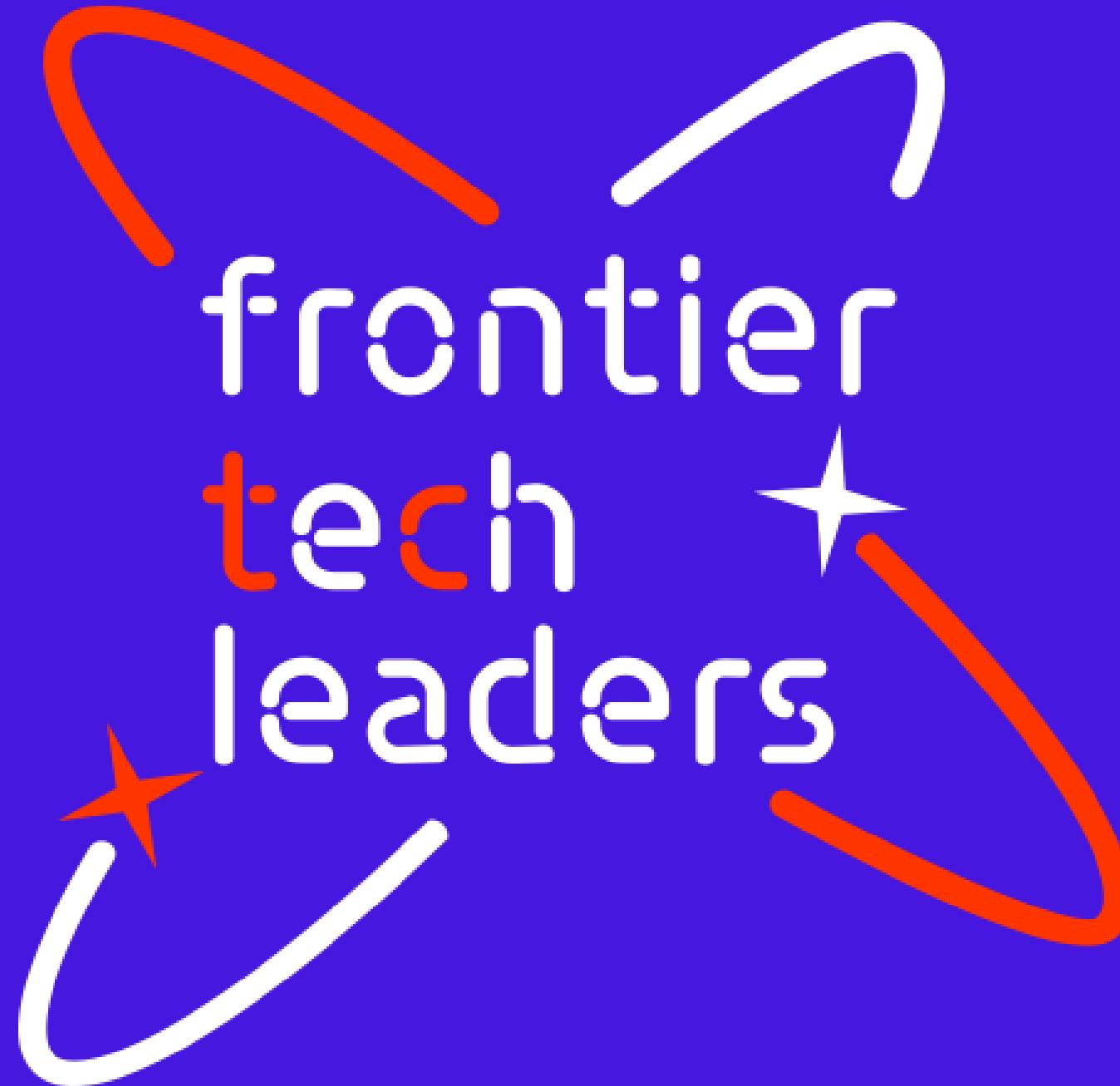
	precision	recall	f1-score
Banded_krait	0.53	0.63	0.58
Bungarus_caeruleus	0.50	0.68	0.58
Bungarus_candidus	0.47	0.26	0.33
Bungarus_fasciatus	0.74	0.65	0.69
Bungarus_lividus	0.80	1.00	0.89
Bungarus_marmoratus	1.00	0.33	0.50
Bungarus_niger	0.85	0.69	0.76
Bungarus_slowinskii	0.00	0.00	0.00
Bungarus_wanghaottingi	0.68	0.57	0.62
Calliophis_bivirgata	0.90	0.60	0.72
Calliophis_macclellandi	1.00	0.36	0.53
Malayopython_reticulatus	0.39	0.50	0.44
Naja_kaouthia	0.68	0.77	0.72
Naja_mandalayensis	1.00	1.00	1.00
Ophiophagus_hannah	0.70	0.47	0.56
Ptyas_korros	0.54	0.83	0.66
Ptyas_mucosa	0.68	0.57	0.62
Python_bivittatus	0.40	0.33	0.36
Python_molurus	0.32	0.57	0.41
...			

- Metrics: accuracy, precision, recall, F1-score (macro and weighted), confusion matrix.

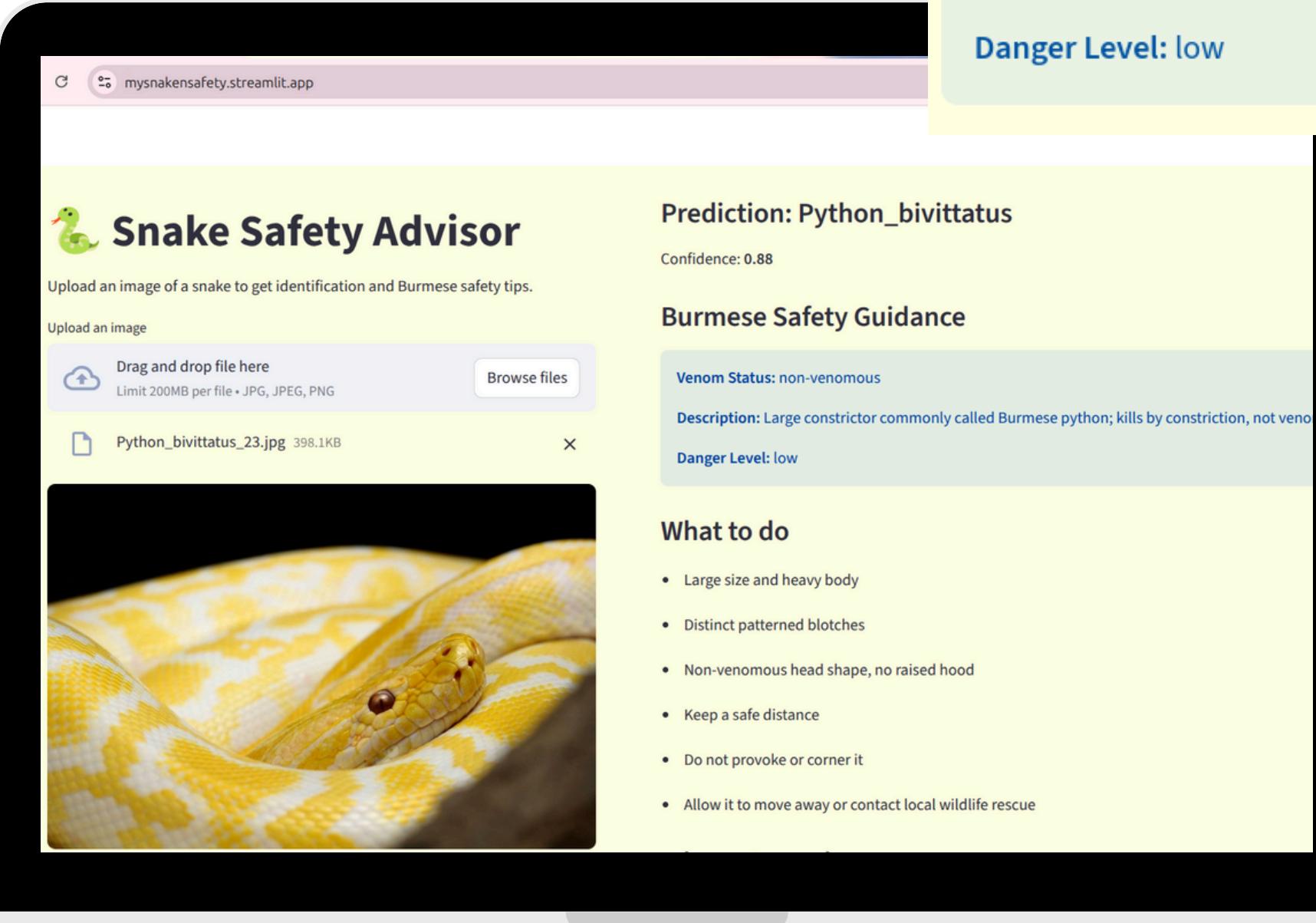
Evaluation Results

- High Accuracy for most species, confirmed by bright yellow cells along the diagonal.
- The model struggles to differentiate species within the *Bungarus* (Kraits) genus.
- Misclassifications are observed between *Python_molurus* and *Python_bivittatus*.





Deployment



Snake Safety Advisor

Upload an image of a snake to get identification and Burmese safety tips.

Upload an image

Drag and drop file here
Limit 200MB per file • JPG, JPEG, PNG

Browse files

Python_bivittatus_23.jpg 398.1KB



Prediction: Python_bivittatus
Confidence: 0.88

Burmese Safety Guidance

Venom Status: non-venomous
Description: Large constrictor commonly called Burmese python; kills by constriction, not venom.
Danger Level: low

What to do

- Large size and heavy body
- Distinct patterned blotches
- Non-venomous head shape, no raised hood
- Keep a safe distance
- Do not provoke or corner it
- Allow it to move away or contact local wildlife rescue

What NOT to do

- Do not attempt to catch or handle
- Do not poke or throw objects at it
- Do not separate from young if found with them
- If bitten: wash wound with soap and water

Apply clean dressing

Venom Status: non-venomous

Description: Large constrictor commonly called Burmese python; kills by constriction, not venom.

Danger Level: low

What to do

- Large size and heavy body
- Distinct patterned blotches
- Non-venomous head shape, no raised hood
- Keep a safe distance
- Do not provoke or corner it
- Allow it to move away or contact local wildlife rescue

What NOT to do

- Do not attempt to catch or handle
- Do not poke or throw objects at it
- Do not separate from young if found with them
- If bitten: wash wound with soap and water

Deployment Overview

- Trained snake ID models exported to ONNX with dynamic batch axes for flexible inference.
- Class metadata (`artifacts/class_map.json`) preserved for consistent label mapping.
- Artifacts enable drop-in integration: HTTP API, batch scripts, or edge inference.
- Supports CPU-only deployment, cloud, on-prem, or embedded edge devices.

API Integration

- POST /predict endpoint accepts image upload (multipart/base64).
- Returns JSON: top-k species + probabilities.
- Stateless server; class mapping ensures consistent outputs.

Security & Monitoring

- Authentication & authorization (API keys / OAuth2).
- TLS for transport security.
- Rate limiting & request size checks.
- Monitor latency, throughput, error rates, confidence distributions.

Challenges & Mitigation



Bungarus Species

Challenges & Mitigation

- Class imbalance → augmentation & sampling.
- Visual features → similarity in Color bands, Head shape, Body texture
- Fine-grained species cues → partial fine-tuning.
- Real-time inference → ONNX + efficient preprocessing.

Conclusion

- Introduces an AI-driven solution for snakebite safety.
- Bridges the gap between research and real-world emergency response.
- Provides fast, reliable, species-based safety instructions.
- Supports SDGs and public health resilience.

Future Work

- Expand species coverage and regional datasets.
- Data preprocessing and augmentation improved generalization.
- Add video-based recognition for live encounters.
- Refine first-aid safety card with richer protocols.
- ONNX export & optional INT8 quantization enables real-world inference.



Thank
you!

frontier
tech
leaders