

AI-Enhanced Equestrian Jump Analysis System

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Agenda

1. Overview
 - a. Dataset
2. Project Scope & Objective
3. Action Recognition Models & Results
4. Auto Annotation Exploration
5. Further work
6. Learnings



Overview

Background: Manual annotation of videos using DartFish software which creates .csv files to run further analytics.

Issues: Prone to human error, inefficient and costly

Ideal solution: Efficient, reliable and robust AI-enhanced system to assist coaches and athletes by tracking and analysing equestrian performances across multiple variables.



Dataset



 Ben MAHER
EXPLOSION W
Breed: KWPN Gelding Sire: Chacco-Blue Age: 13



1.7

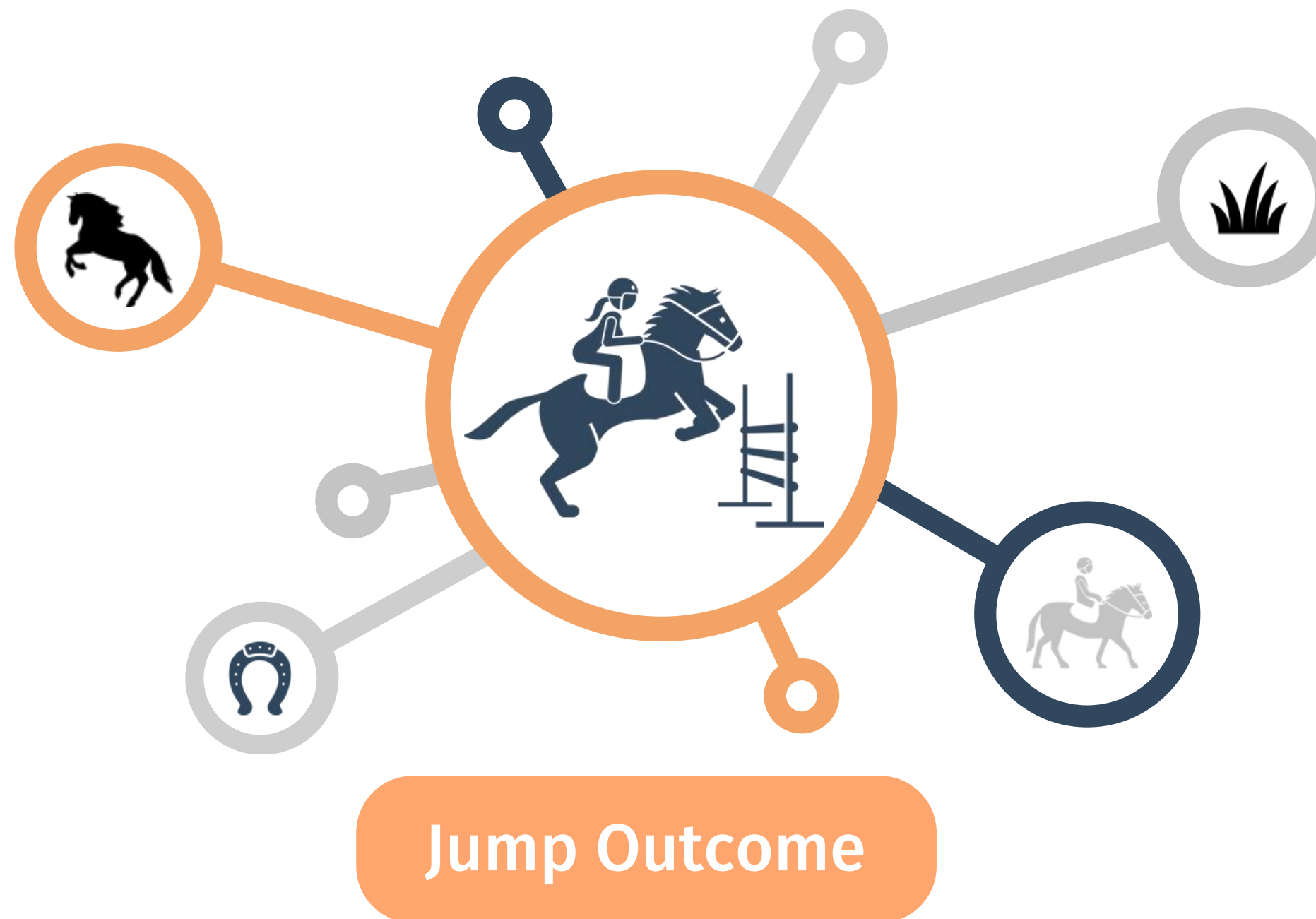
Project Scope & Objective

Rein

On which side the horse leans

Leg

Takeoff and landing leg for each jump



Surface

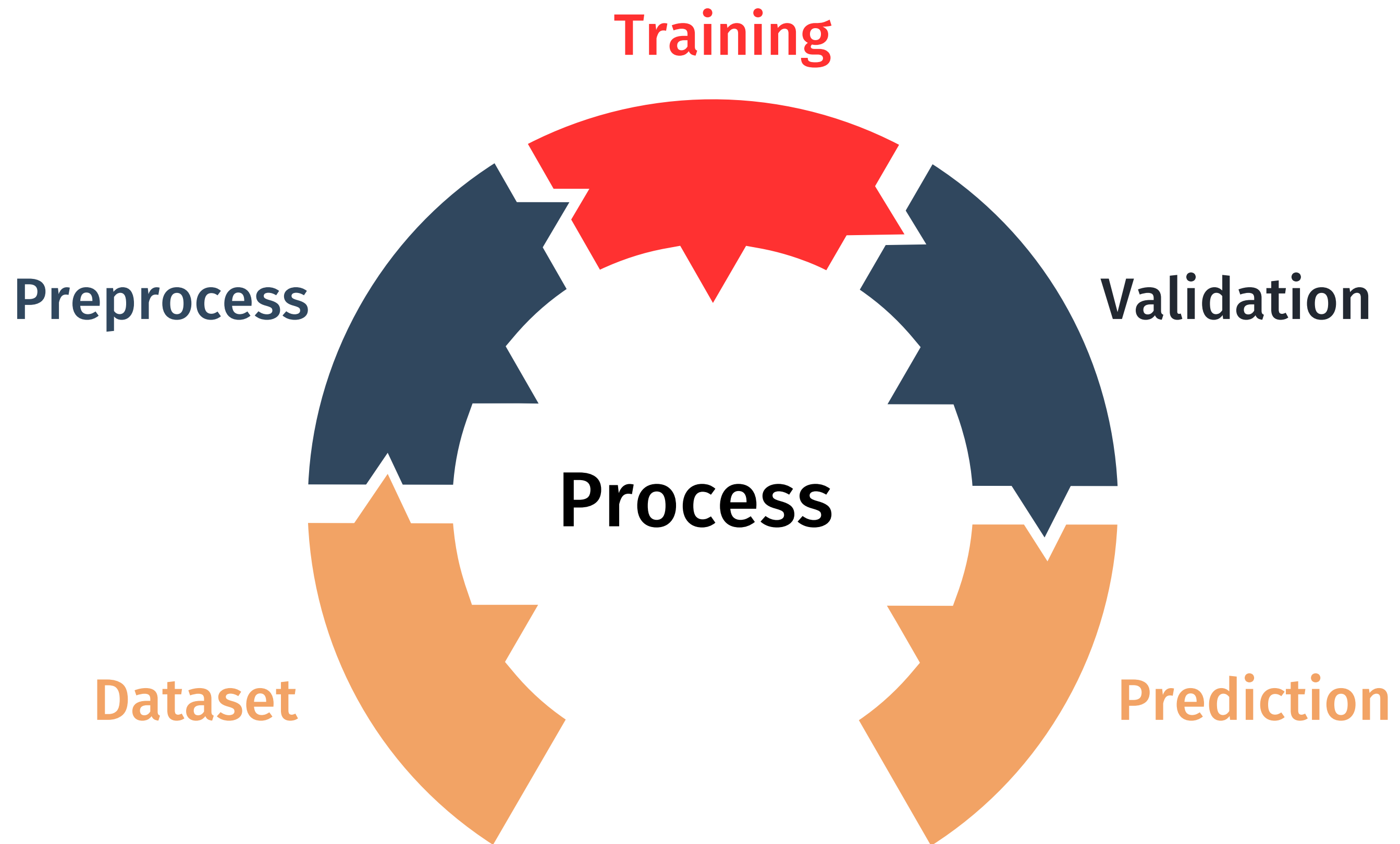
Either sand or grass

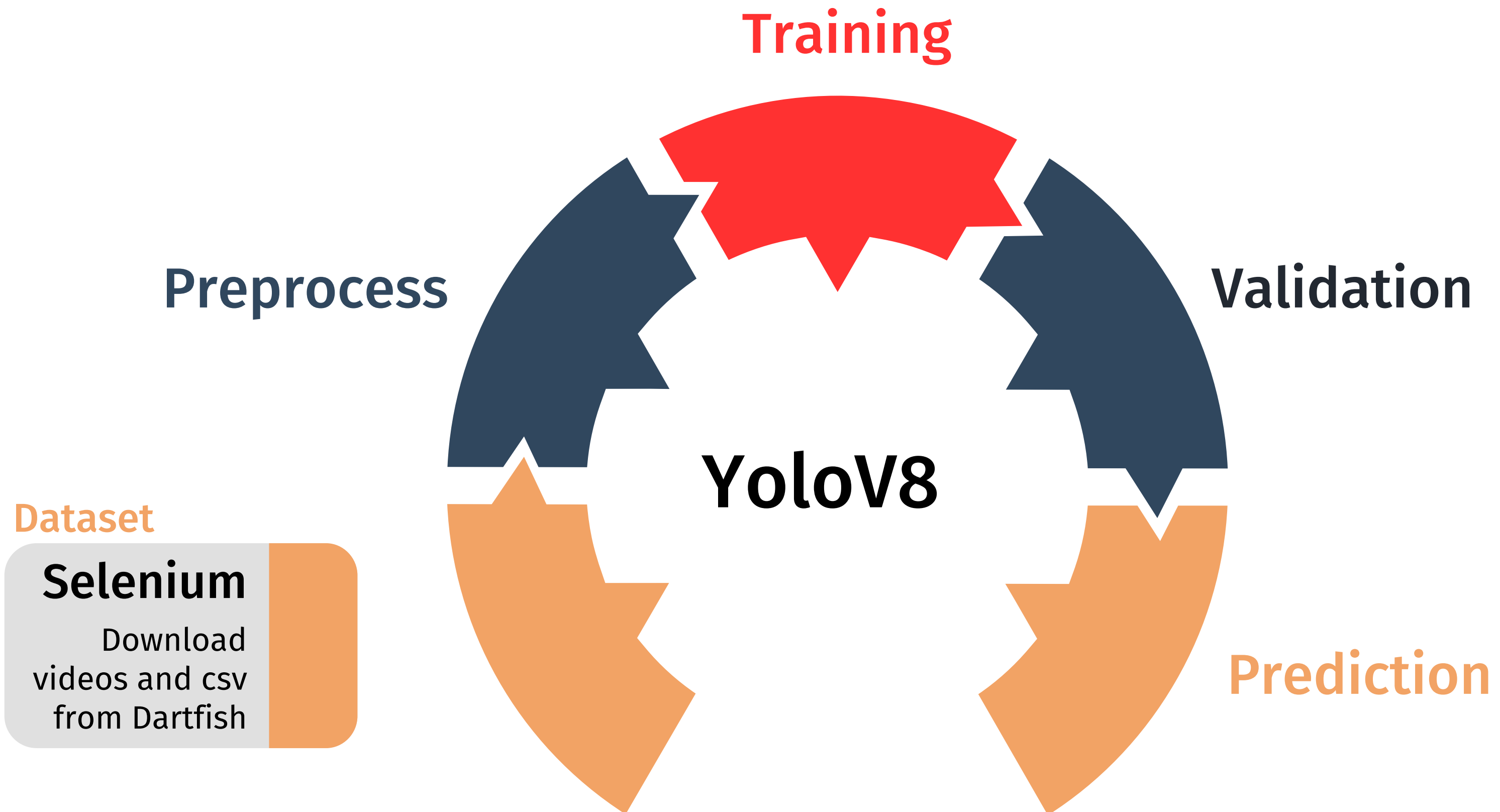
Jump Type

Parallel & upright

Strides

Number of strides between jumps





Preprocess

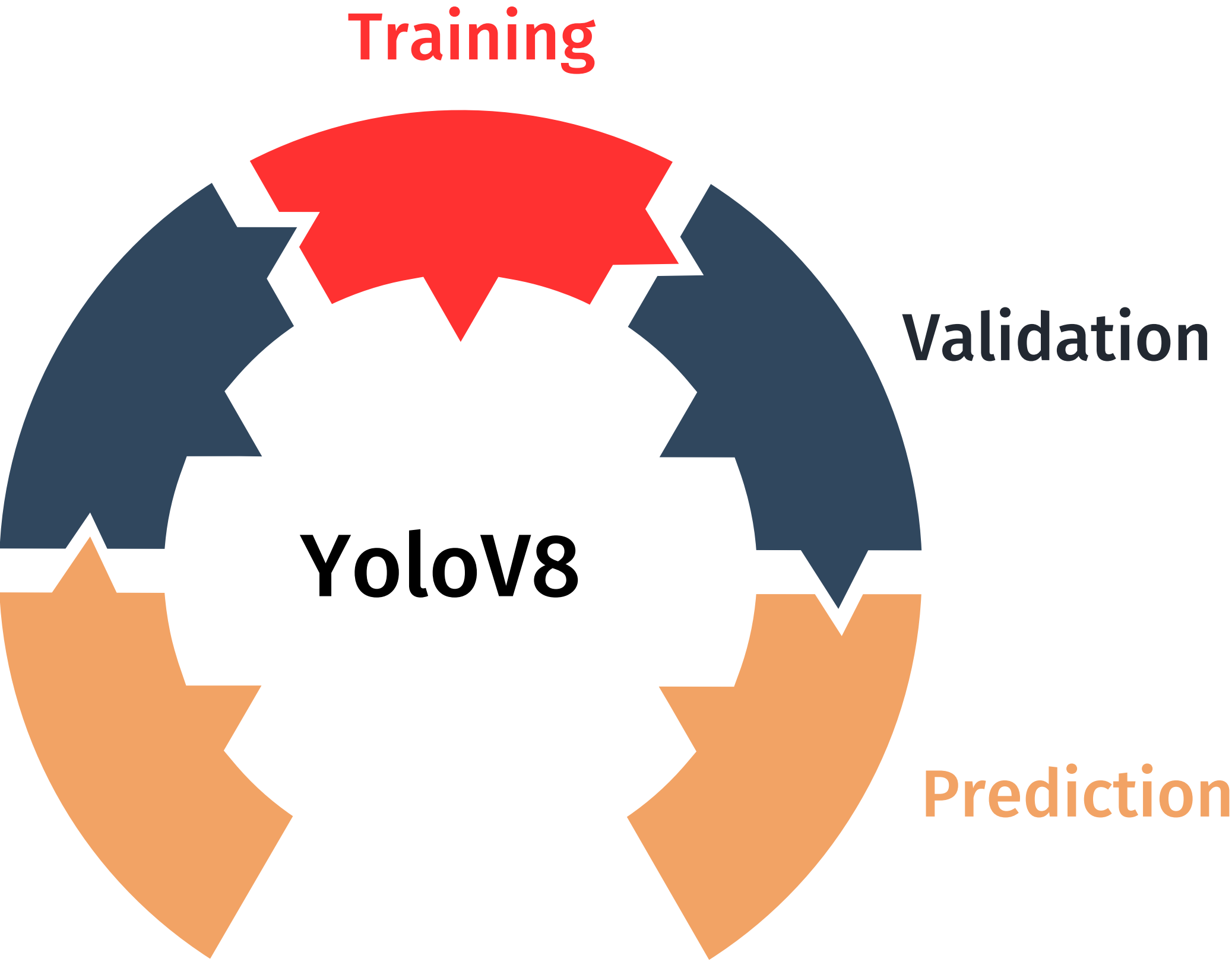
Extract Jumps

Split each video
into clear, unclear
and running

Dataset

Selenium

Download
videos and csv
from Dartfish



Trained

YoloV8

Runs only for
human and
horse classes

Preprocess

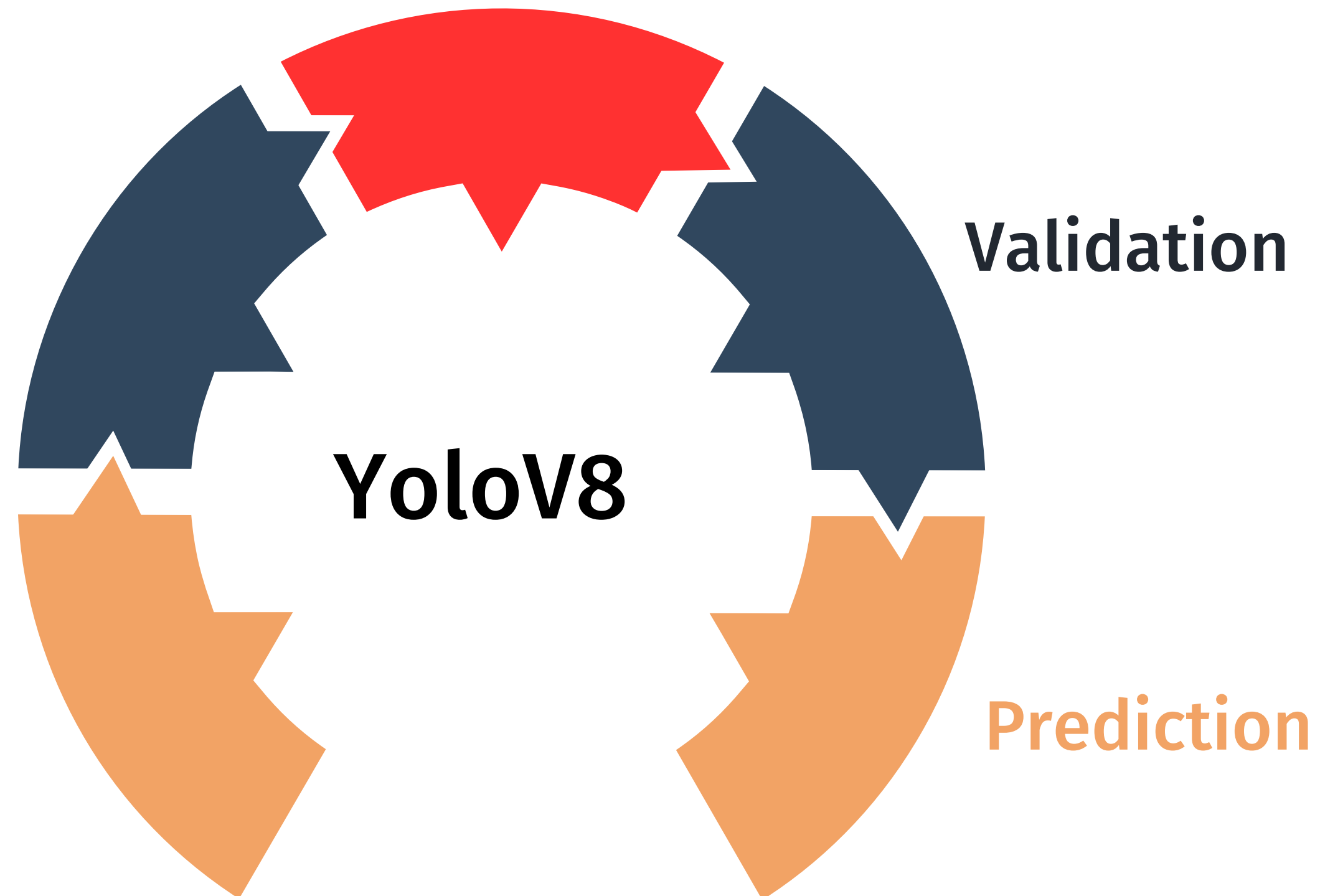
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Validation

For Accuracy

The ratio of
predicted output
to the real data

Prediction

Segmentation

Got rider and the
horse segmented
as output

YoloV8

YoloV8 Result



Objective



Unsuccessful Jump

Exploring Action Recognition

Results

Training

MMAction2

On carefully
selected videos with
suitable angles

Validation

For Accuracy

The ratio of
predicted output
to the real data

Prediction

Jump Outcome

Classifications using
two classes weather
clear or unclear

Preprocess

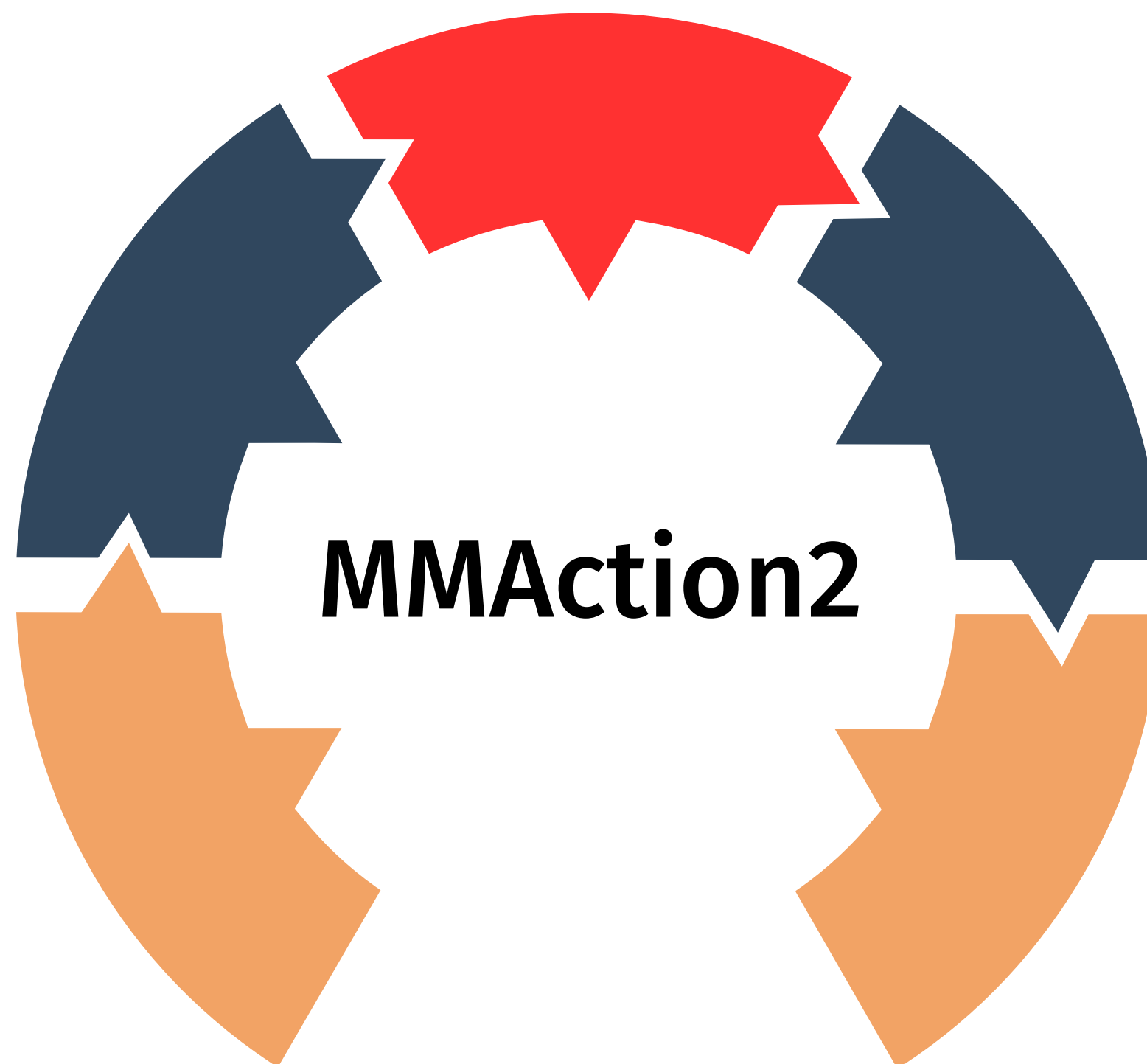
Technique

Implementation
and integration in
the process.

Dataset

Preparation

Place videos in
separate folders
with labeled csv



Training

MMAction2

On carefully
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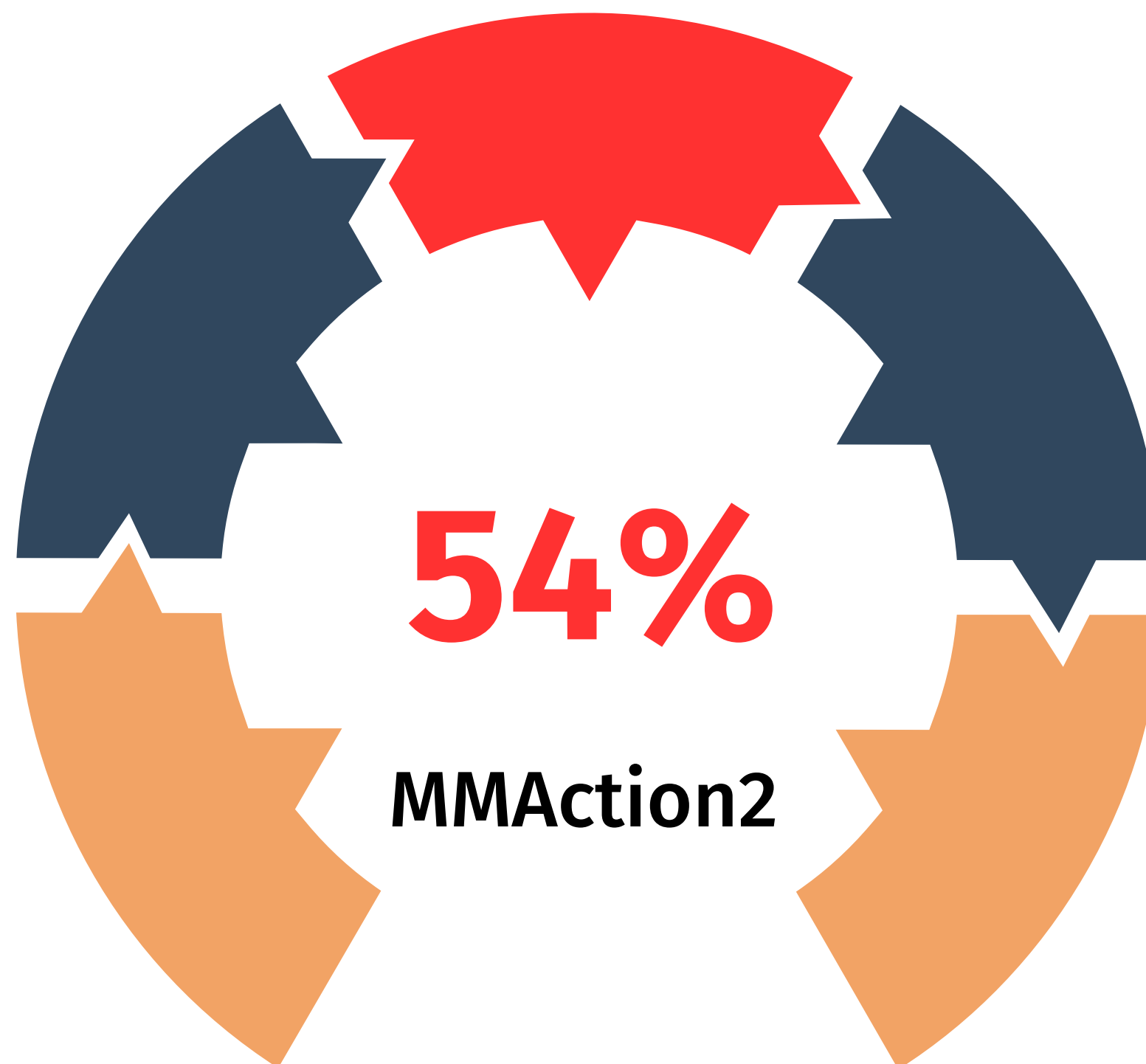
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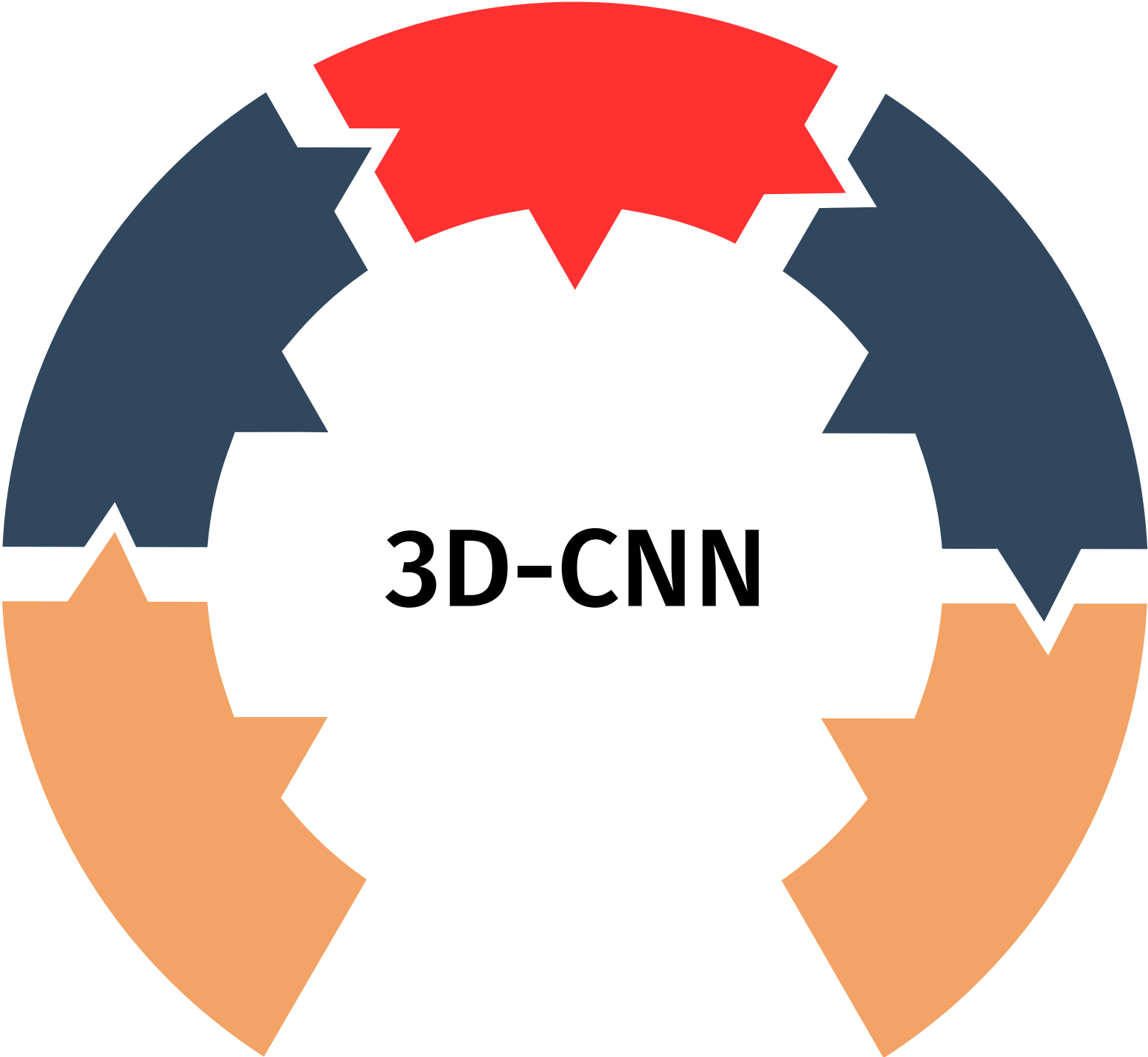
3D-CNN
50 epochs on 300+
jumps

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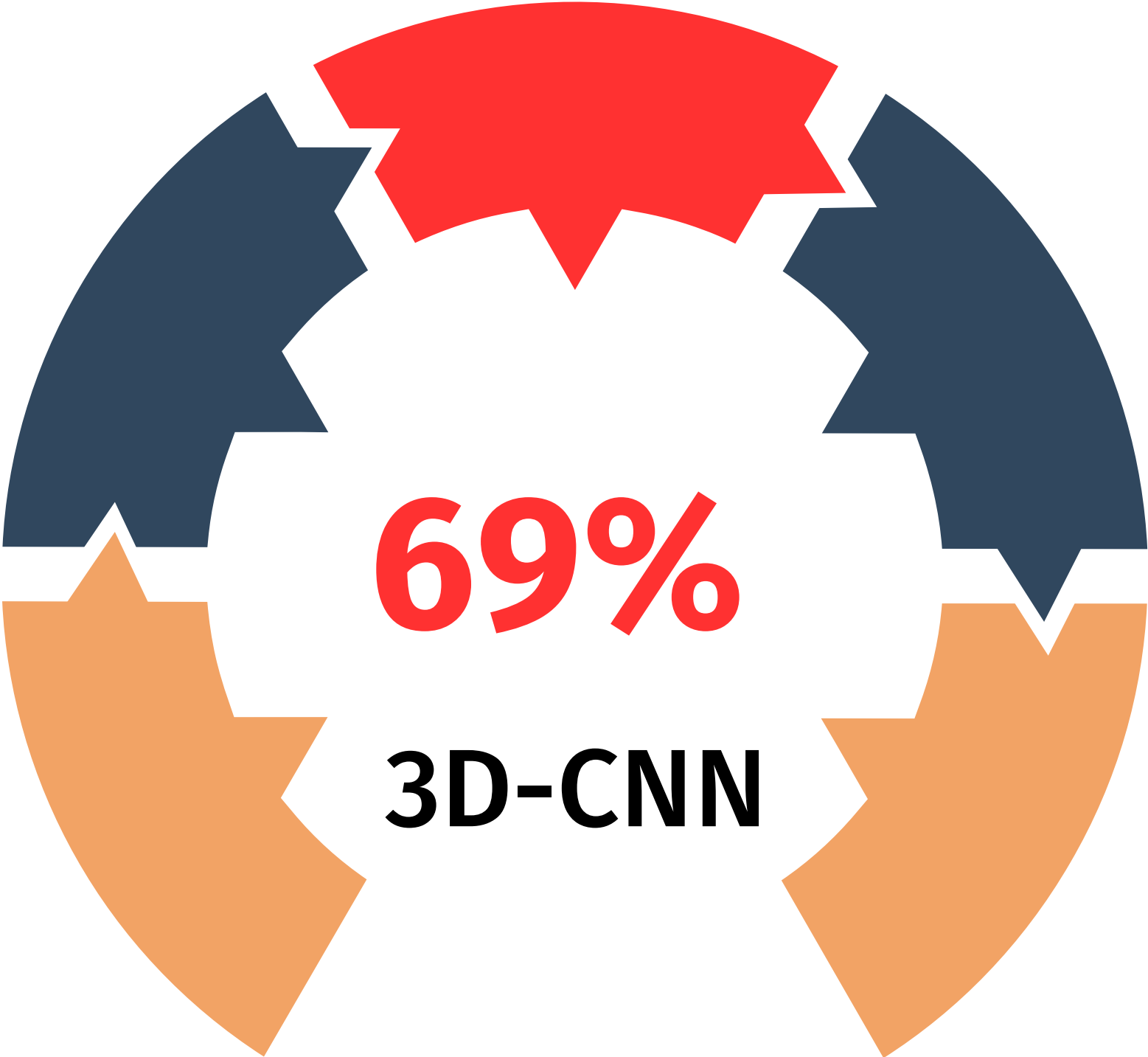
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3D-CNN
MMAction2

Preprocess

Technique
Implementation and integration in the process.

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Training

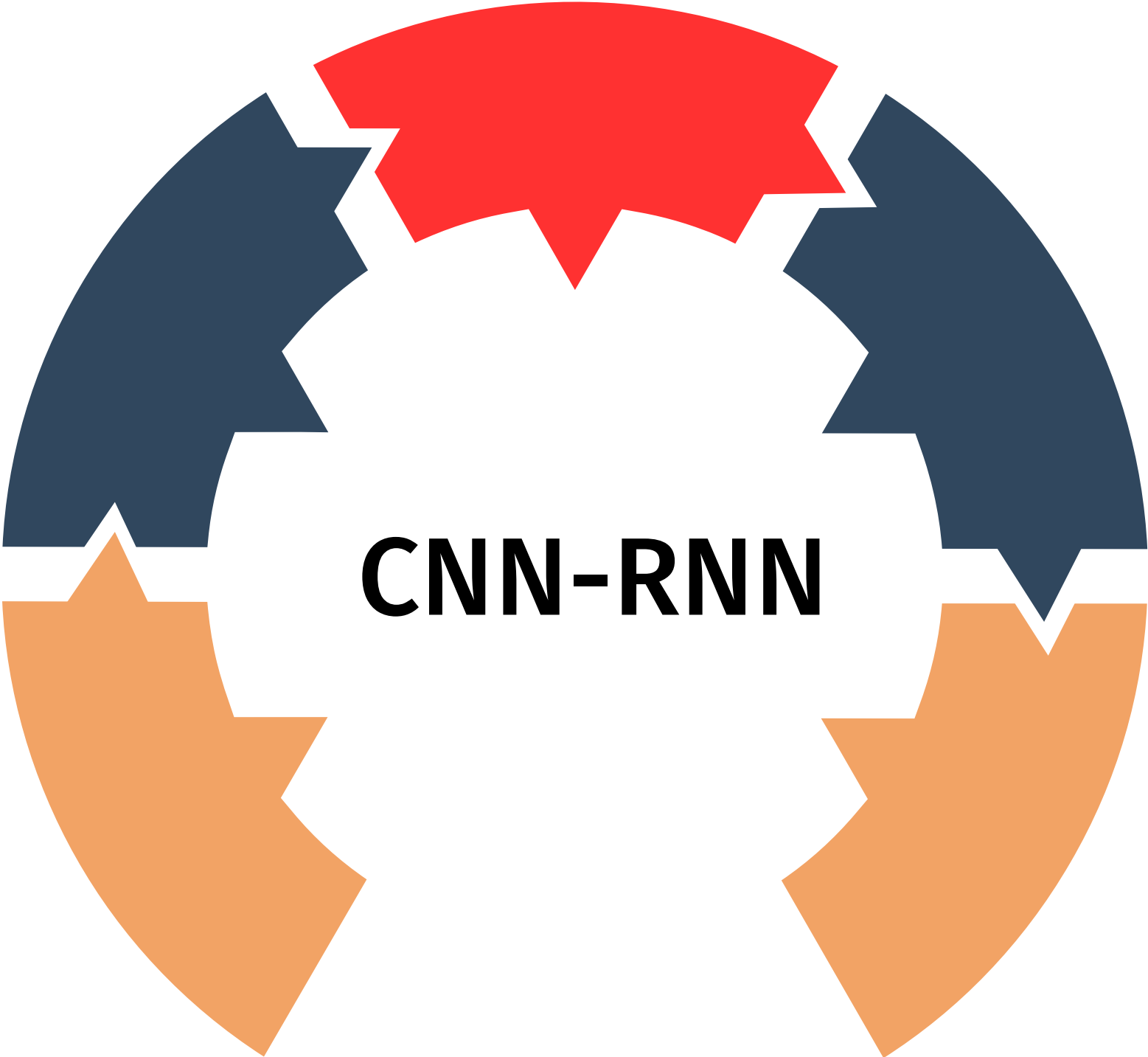
CNN-RNN
50 epochs on 300+ jumps

Validation

For Accuracy
The ratio of predicted output to the real data

Prediction

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Classifications using two classes weather clear or unclear



3D-CNN
MMAction2

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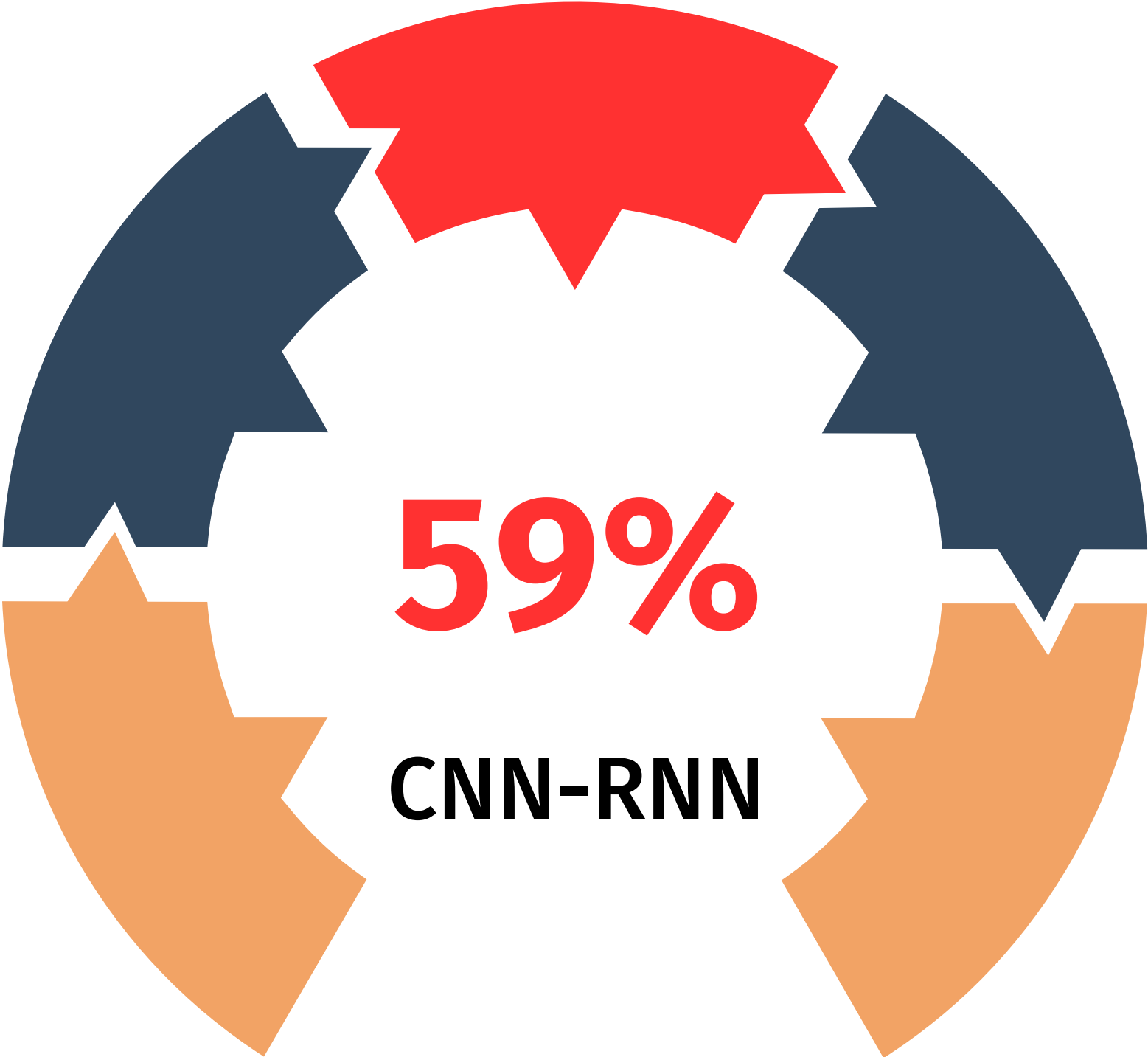
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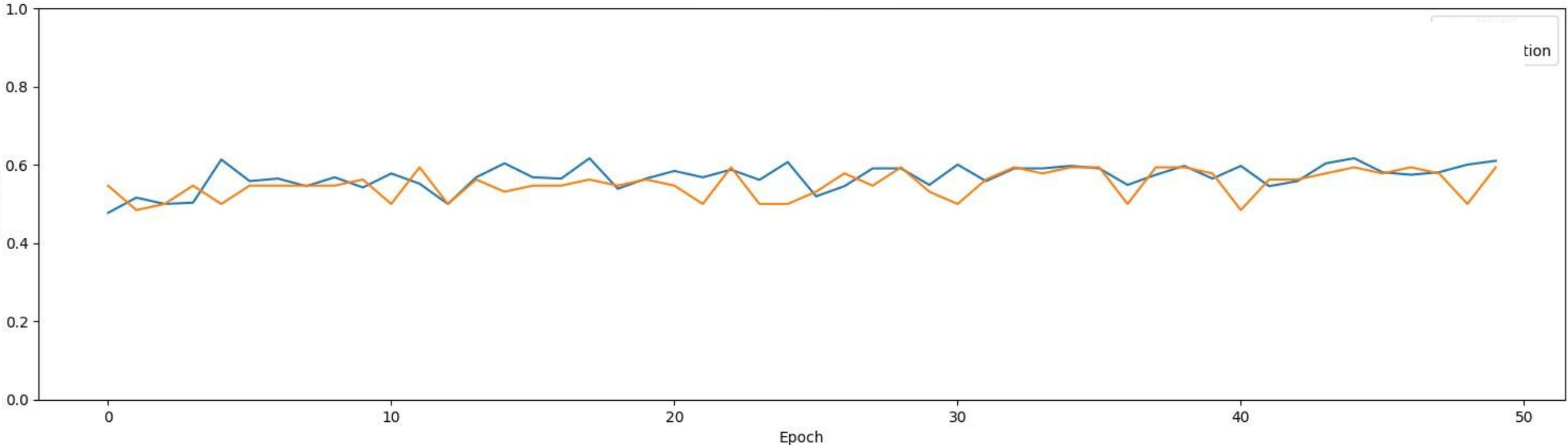
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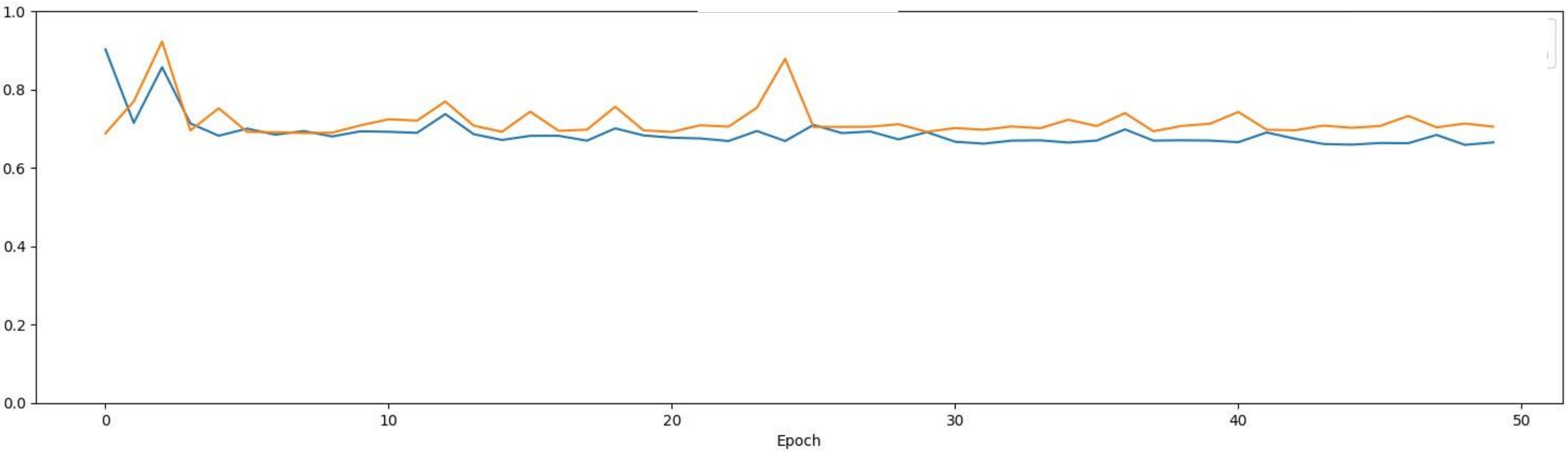
3D-CNN Result

Train
Validation

Accuracy

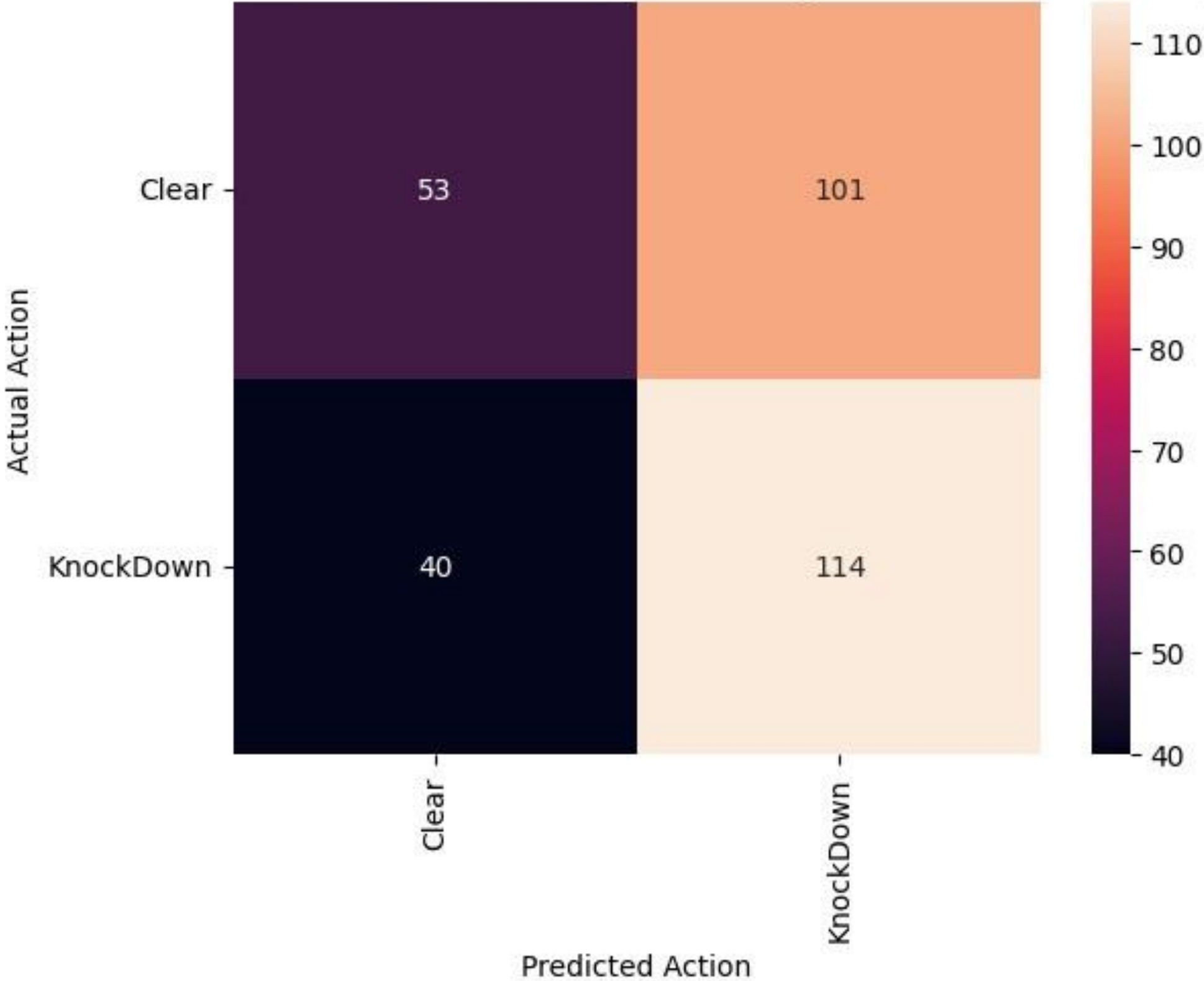


Loss

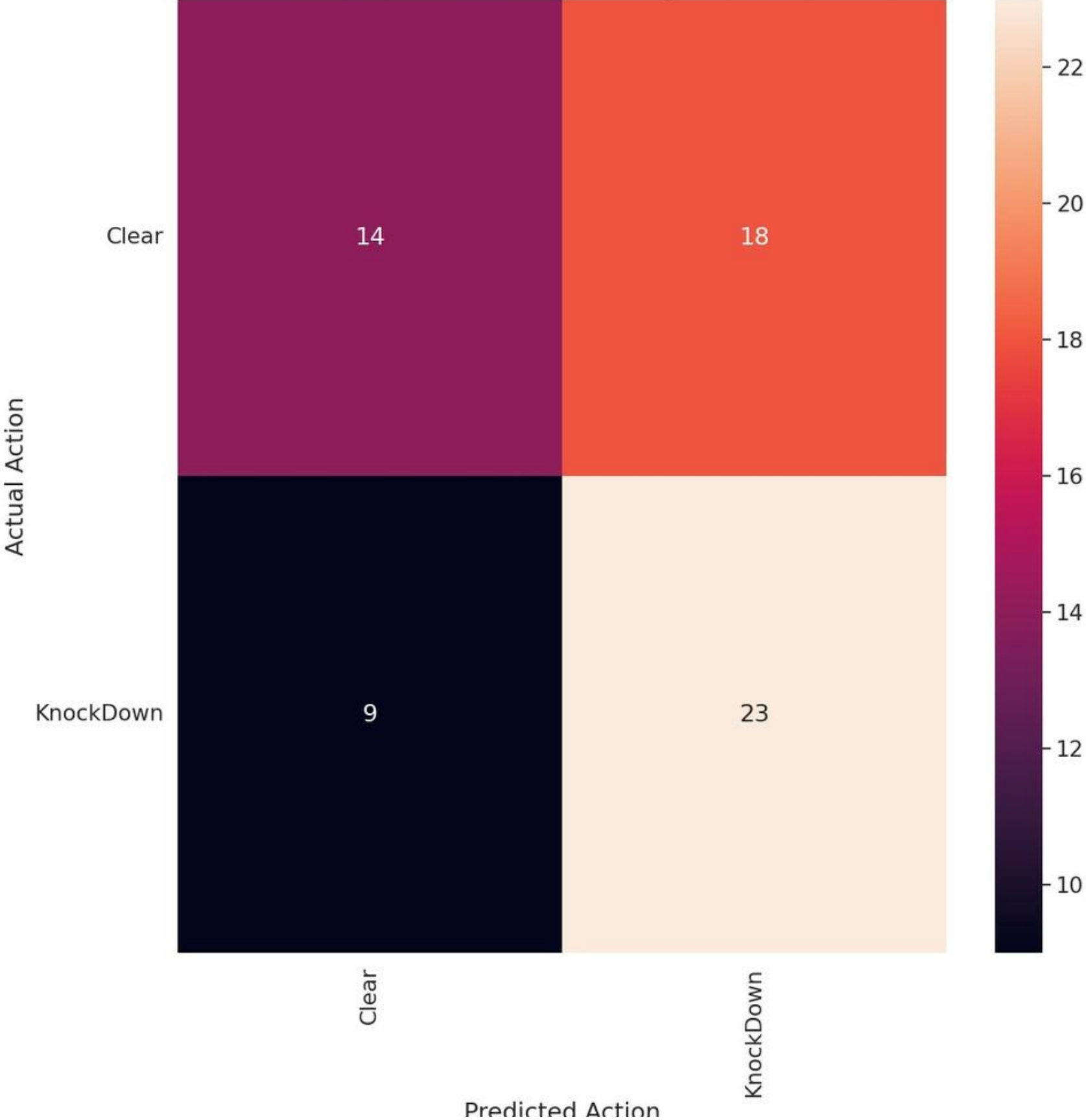


3D-CNN Result

Confusion matrix of action recognition for training

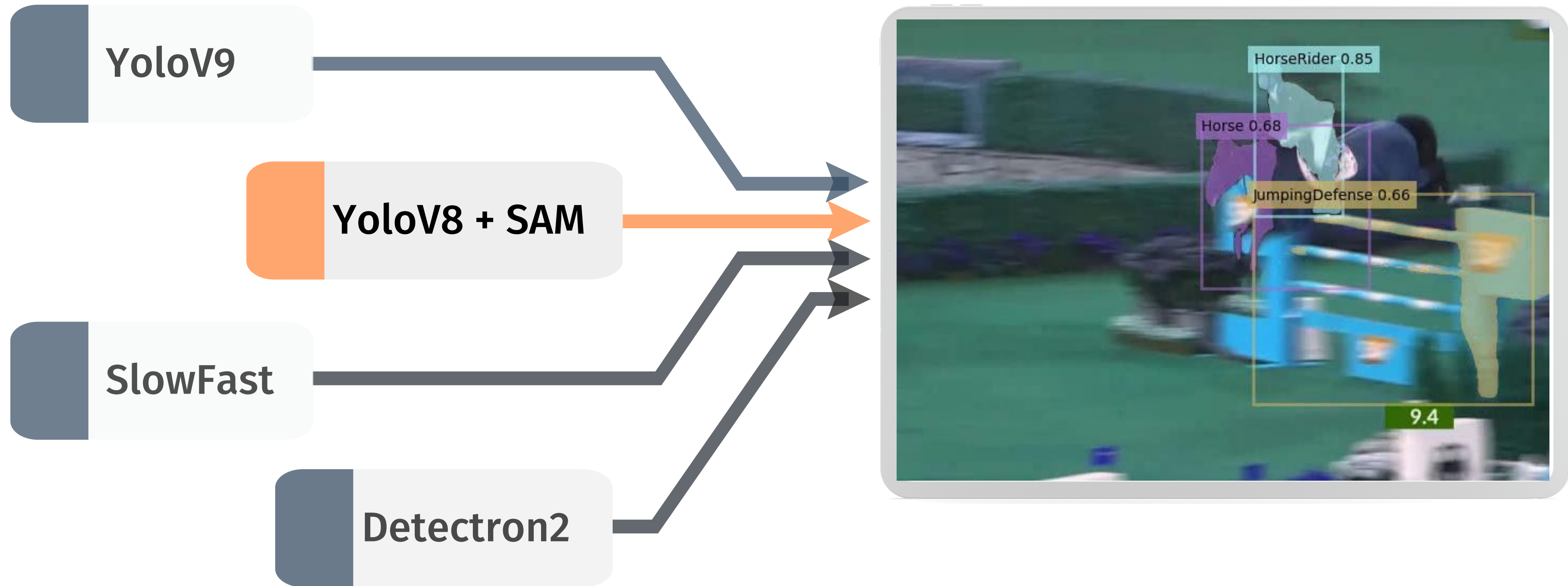


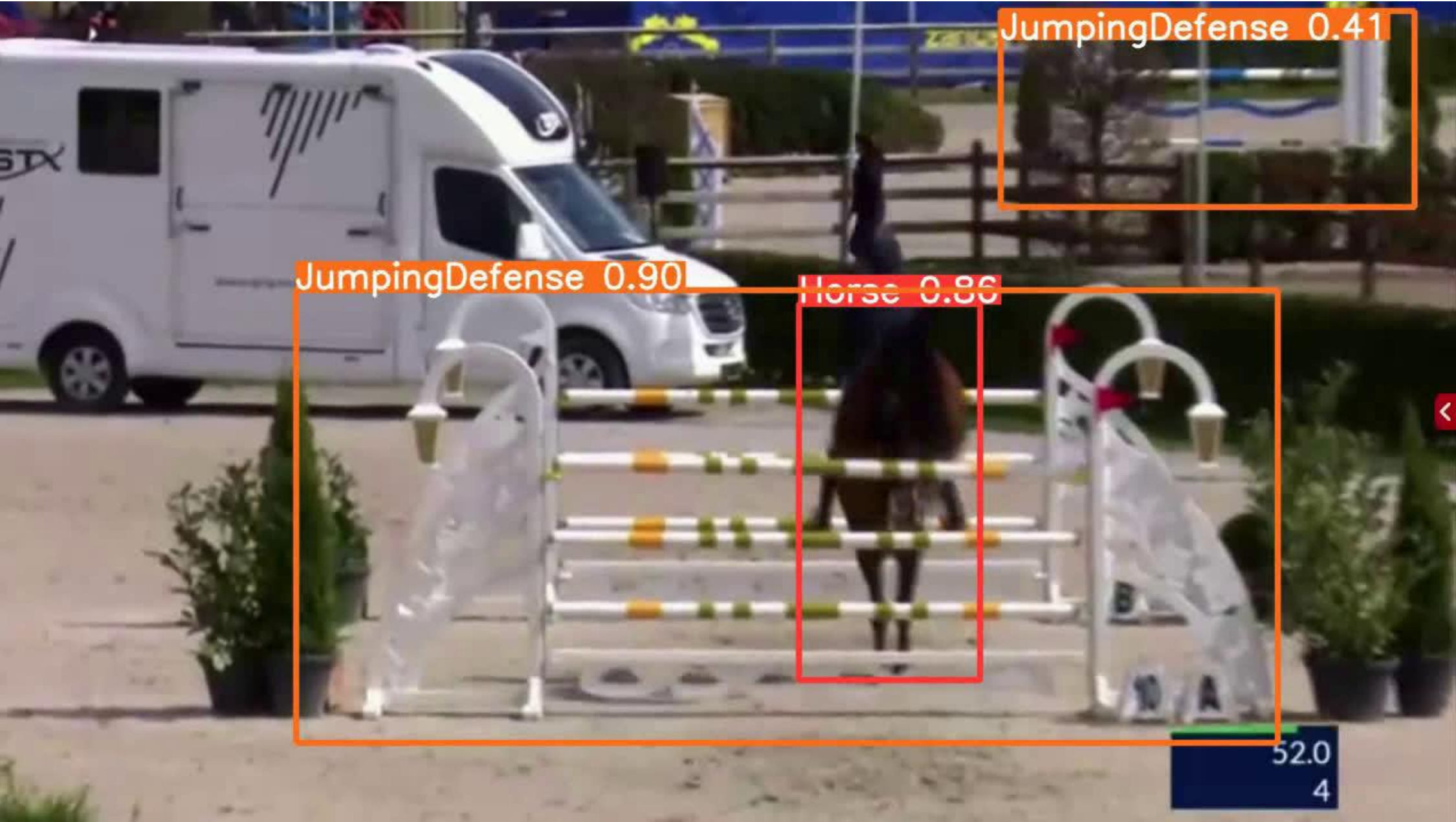
Confusion matrix of action recognition for test



Auto Annotation

Auto Annotation



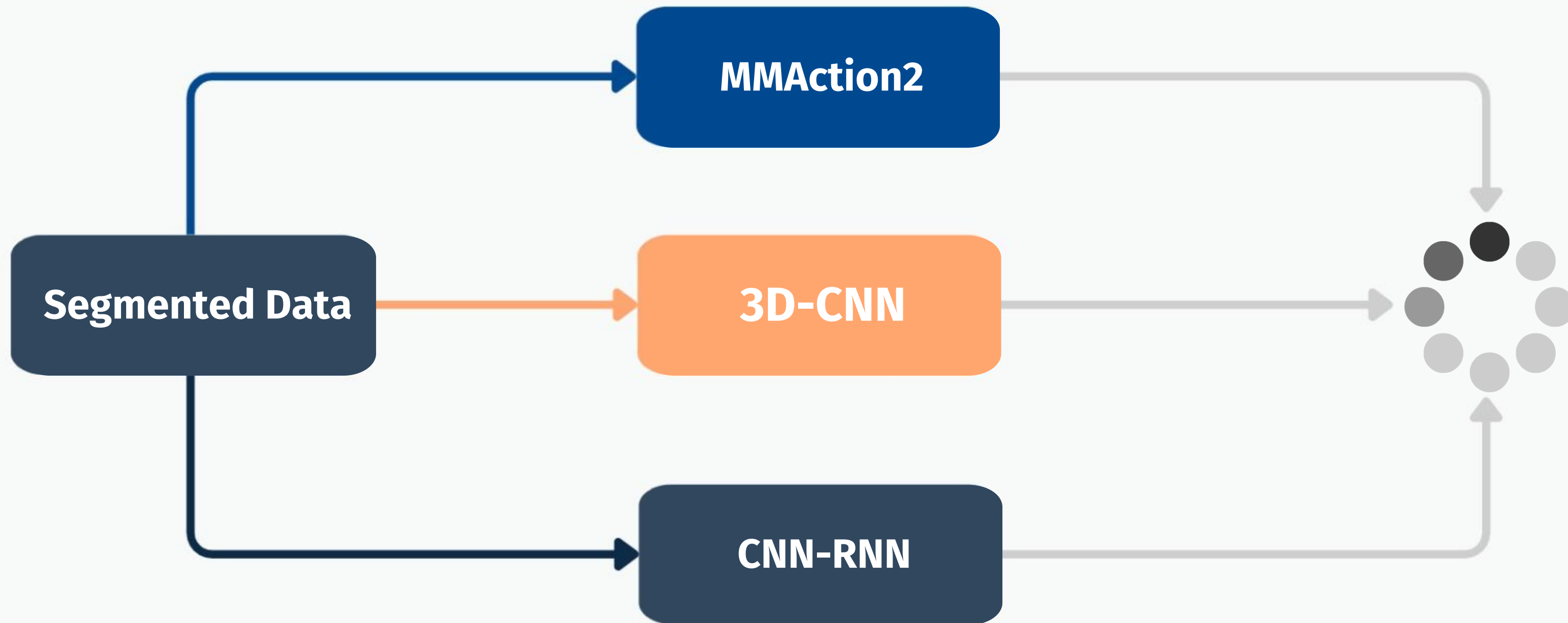


YoloV8

Segment Anything Model (SAM)



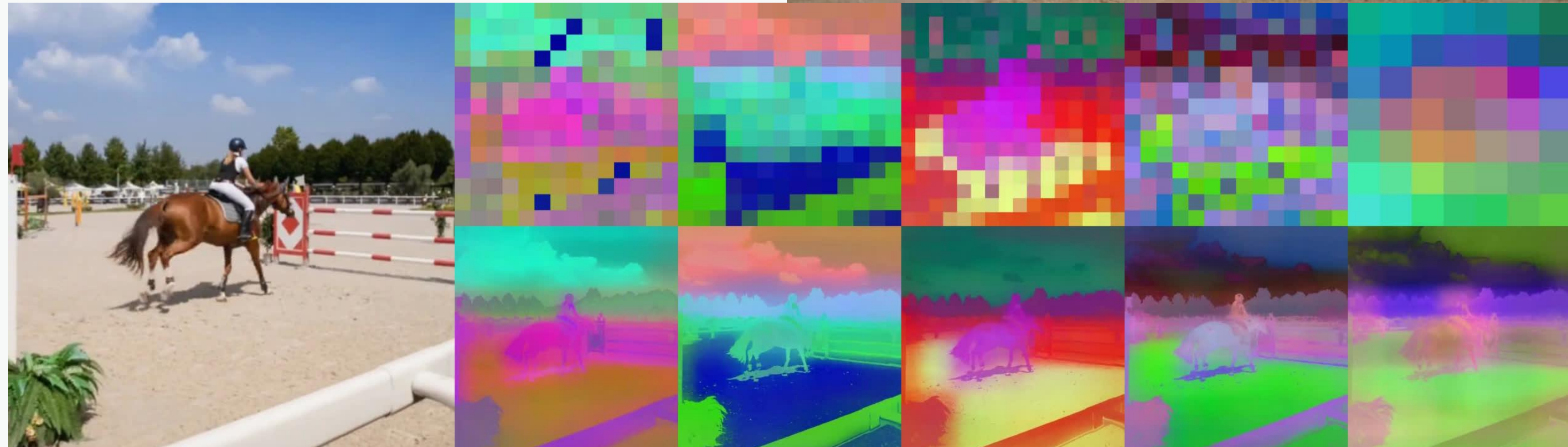
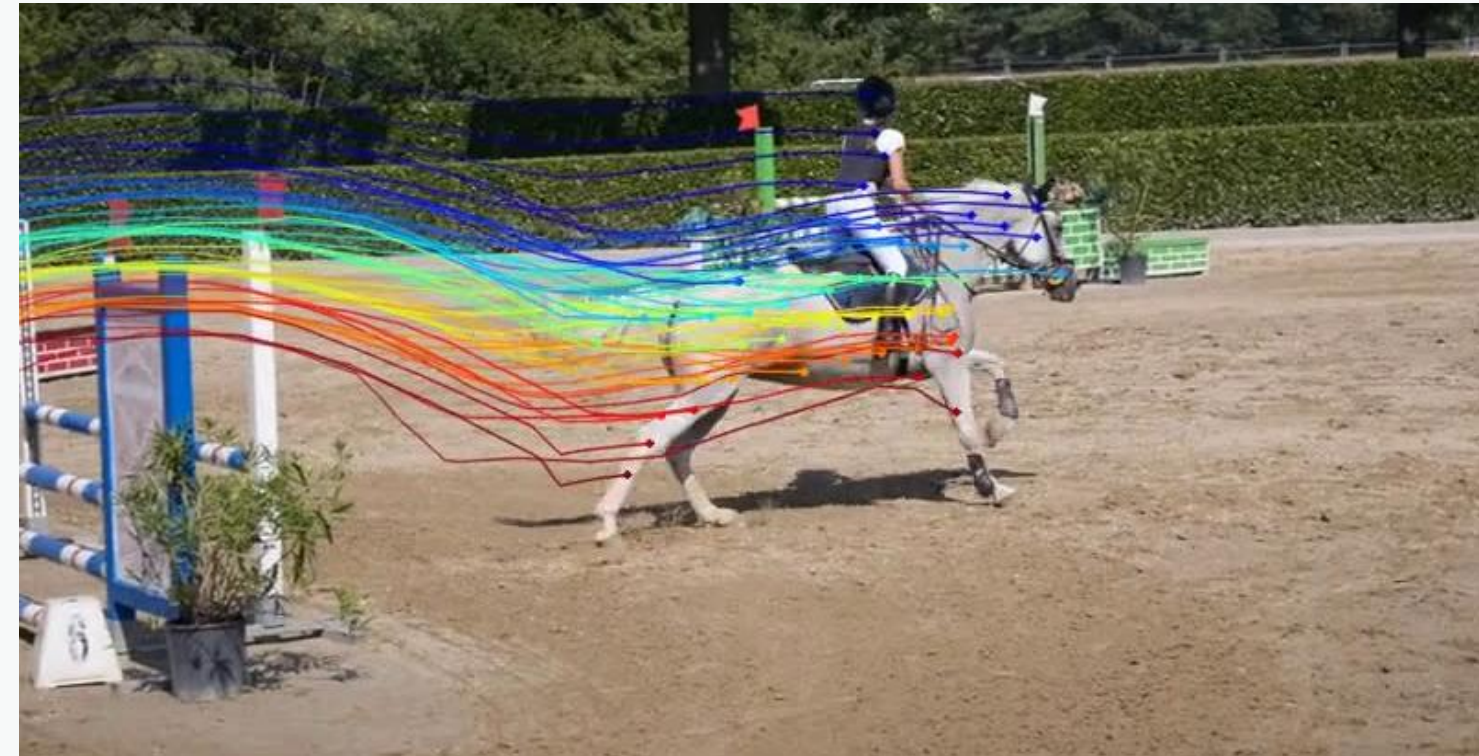
Further Work (before submission)



Future Work

OmniMotion: Tracking Everything Everywhere All at Once (2023)

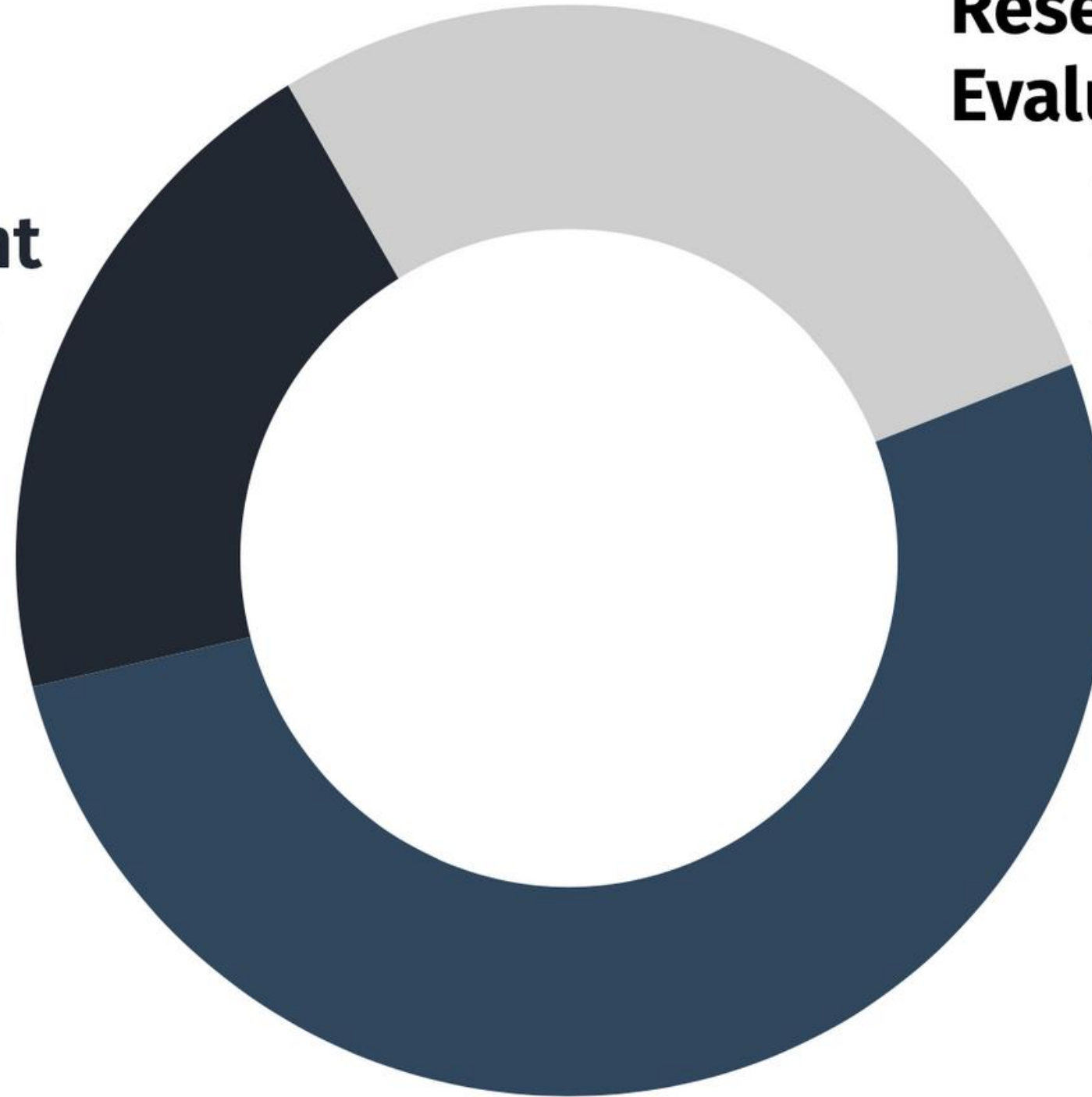
FeatUp: A Model-Agnostic Framework for Features at Any Resolution (2024)



Learnings

Project Management

- Planning & Scrum records
- Dashboards
- Documentation & Communication



Research, Monitoring & Evaluation

- AI Models' Research
- Technical Problem Solving
- Goals/Outcome/Output

Technical Acumen

- Programming/Debugging
- Azure/Linux/Colab/Jupyter
- Programming Documentation

Thank You
