

AI4AS | Monday June 29th



Testing synthetic data integration for automated fish detection on beam trawl vessels

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Why synthetic data for fisheries monitoring?

- Increasing importance of **automated catch monitoring** for **scalable and reliable fisheries data collection**
- Need for **diverse, well-labelled data**
- **Real data imagery is challenging** in trawl fisheries:
 - Mixed catches
 - Overlapping individuals
 - Variable lighting
 - Rare or protected species
 - High annotation effort

Synthetic data could help **fill these gaps** - if it **transfers reliably to real imagery**



From synthetic data promise to workflow sensitivity

Main question:

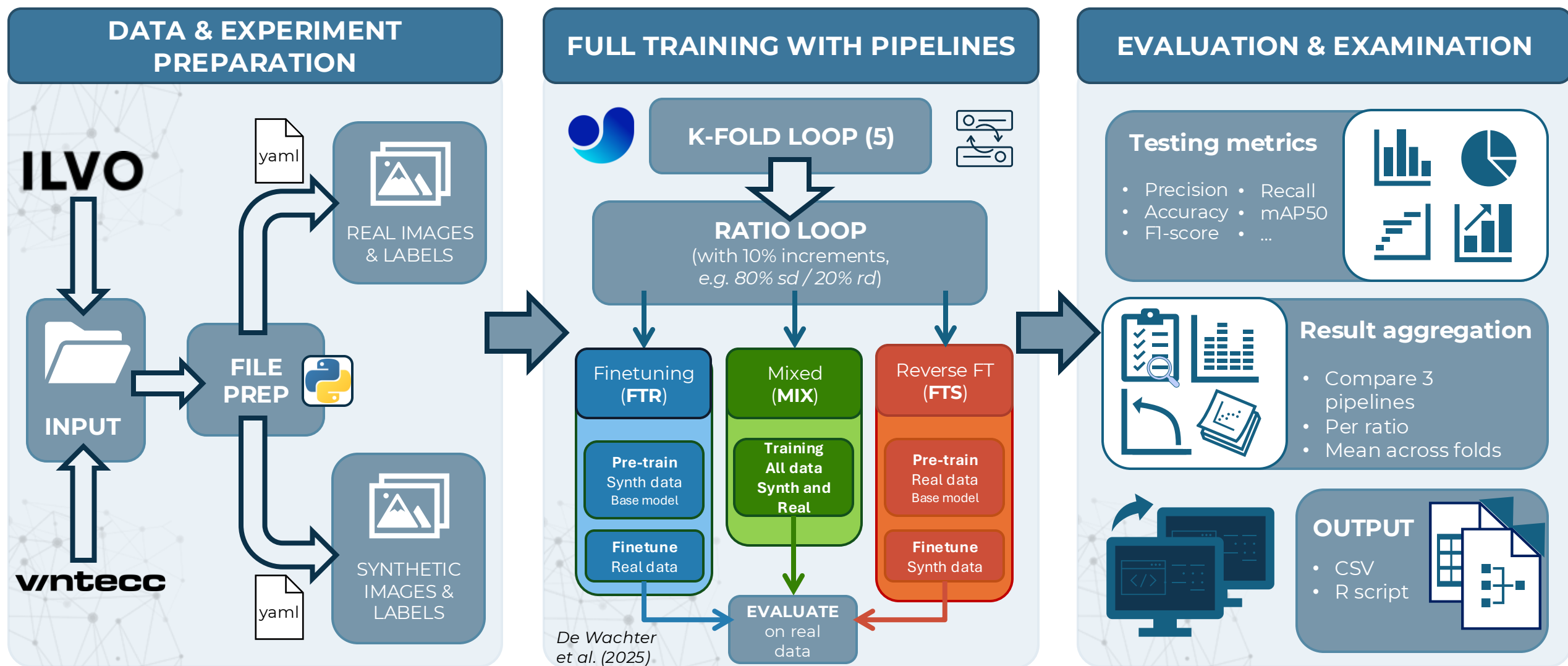
Can **synthetic fish** imagery **improve detection models** for real fisheries monitoring, and which **workflow choices** determine whether this **improvement is reliable**?

We evaluate sensitivity to:

- 1 Synthetic–real training strategy
- 2 Synthetic–real data ratio
- 3 Annotation and preprocessing workflow
- 4 Biological and visual diversity

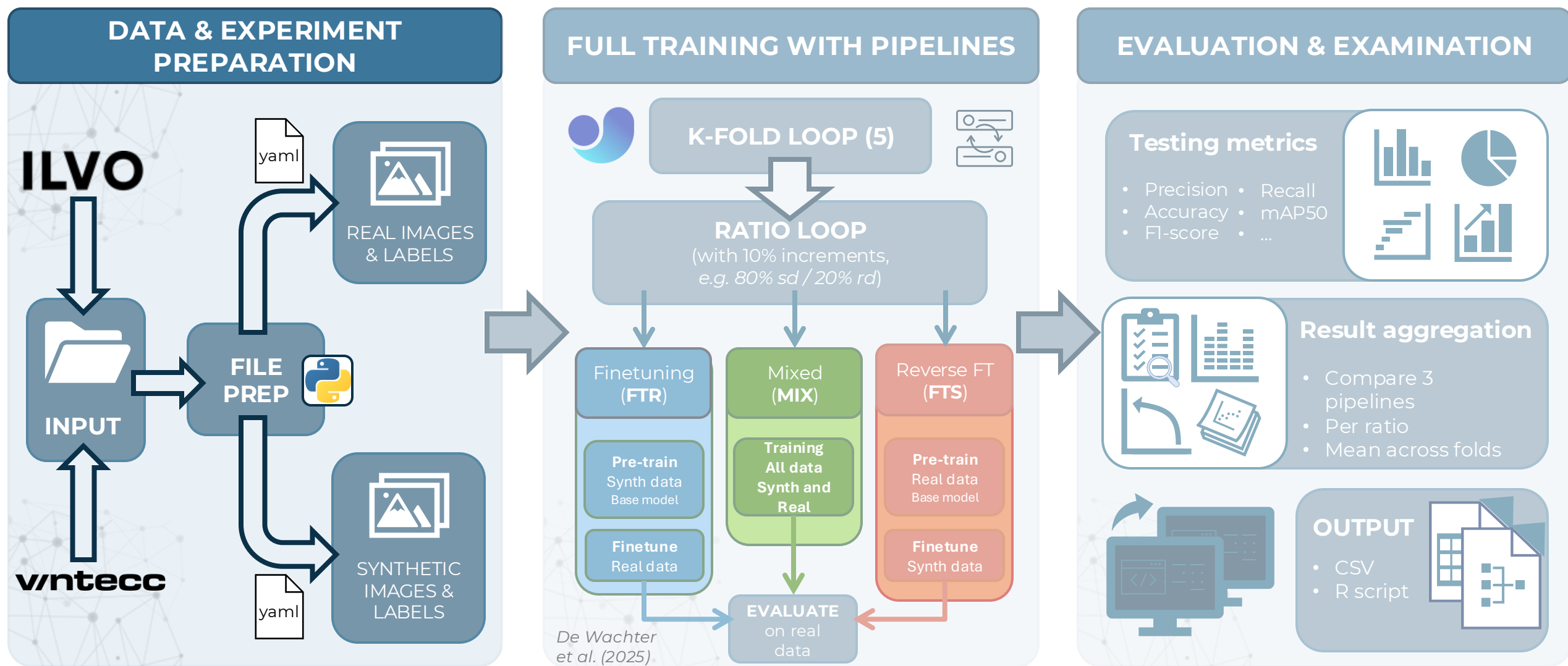
Project workflow

detection & segmentation model training



Project workflow

detection & segmentation model training



Project workflow

Real data

Controlled lab scenario's

- Two **species**: brill or plaice



Project workflow

Real data

Controlled lab scenario's

- Two **species**: brill or plaice
- **Orientation**:
 - belly, back, or mixed



back

mixed

belly

Project workflow

Real data

Controlled lab scenario's

- Two **species**: brill or plaice
- **Orientation**:
 - belly, back, or mixed
- **Overlap**:
 - none, moderate, or heavy



none

moderate

heavy

Project workflow

Real data

Controlled lab scenario's

- Two **species**: brill or plaice
- **Orientation**:
 - belly, back, or mixed
- **Overlap**:
 - none, moderate, or heavy
- **Illumination**:
 - moderate, normal, or strong



moderate

normal

strong

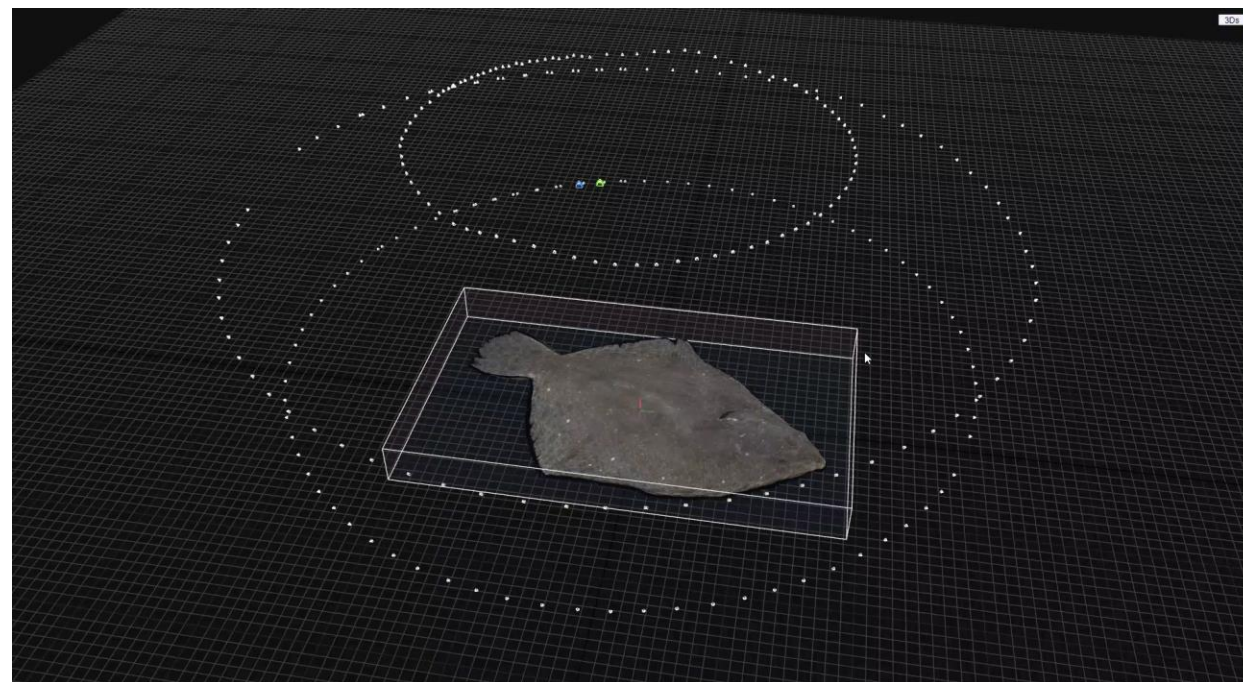
Project workflow

Synthetic data

vntecc

Created by Vintecc

- Photogrammetry-based 3D fish models
- Rendered under varying pose, lighting, and backgrounds
- Automatically labelled during generation

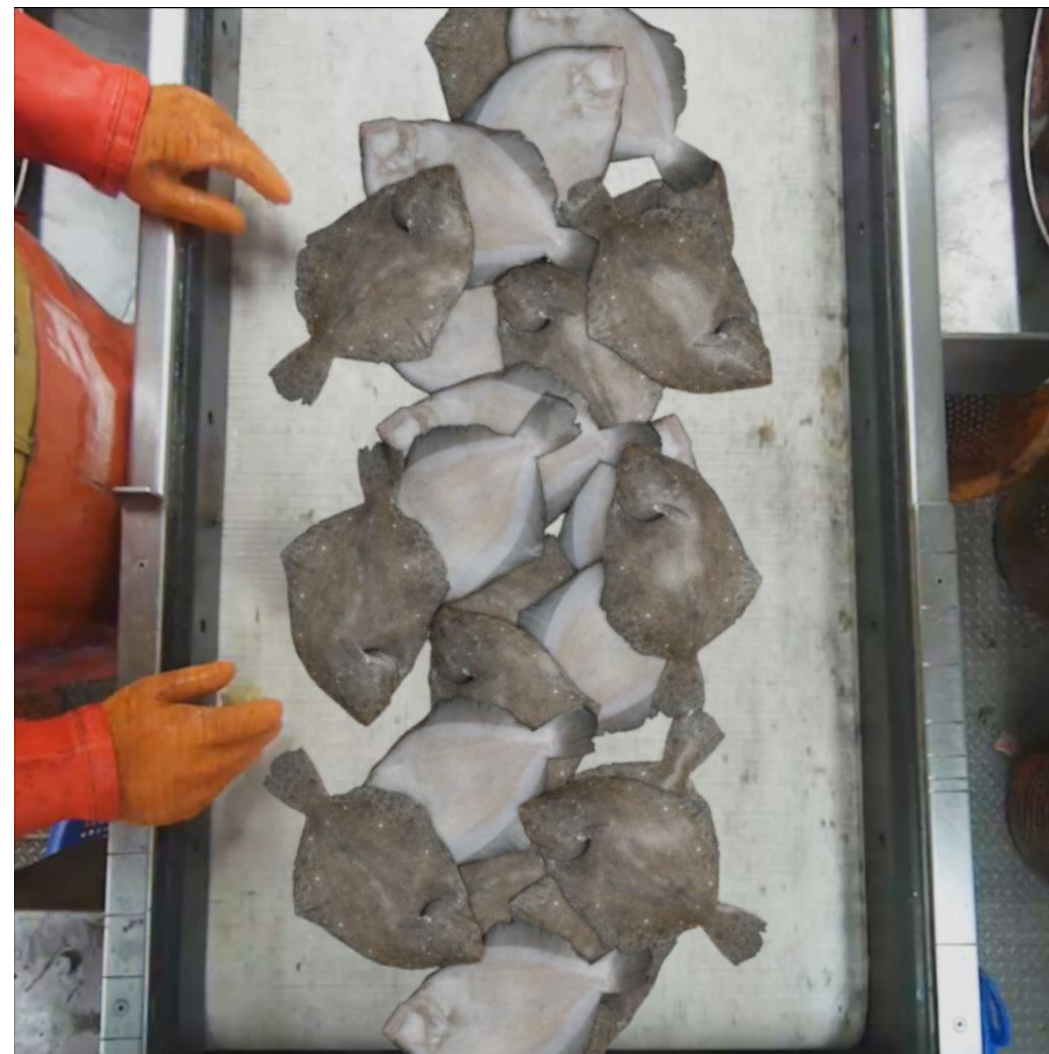


Project workflow

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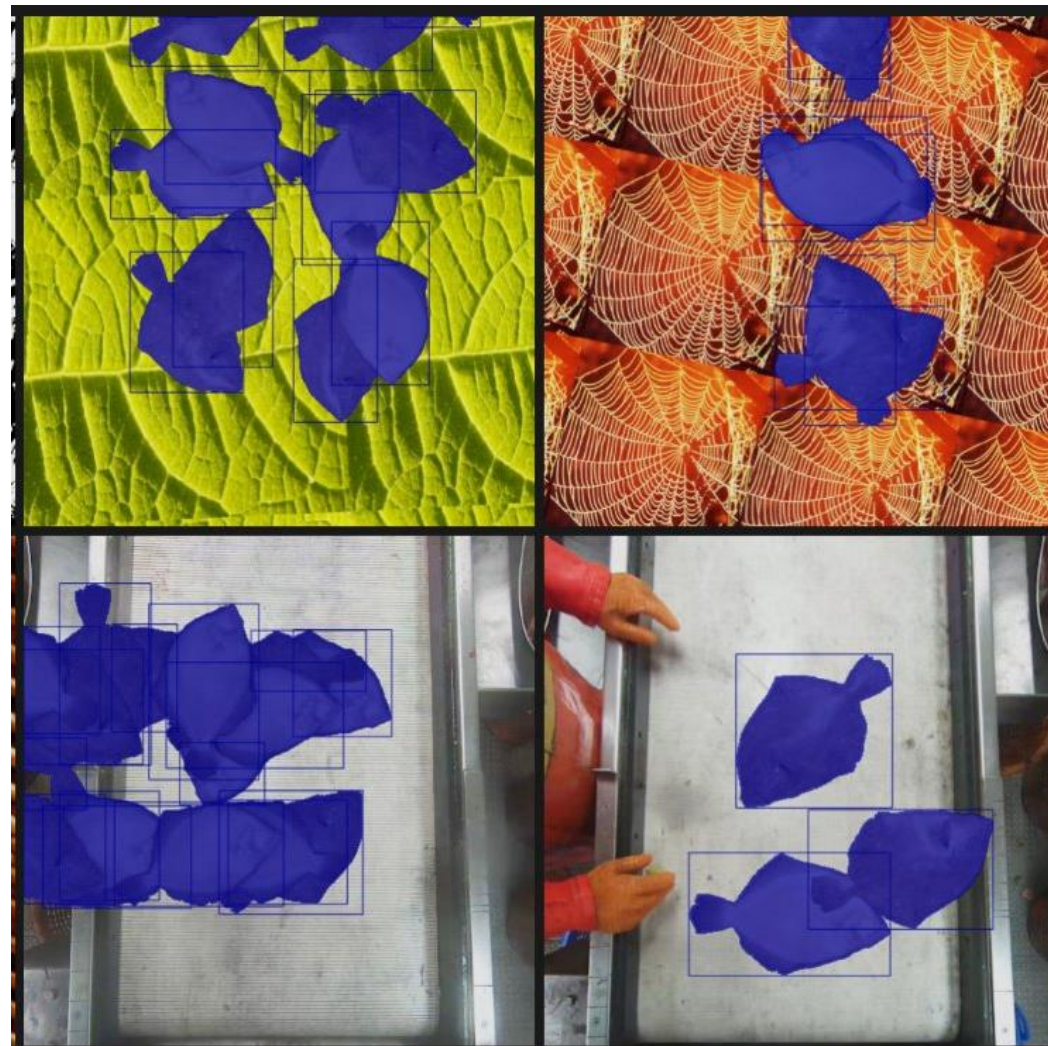


Project workflow

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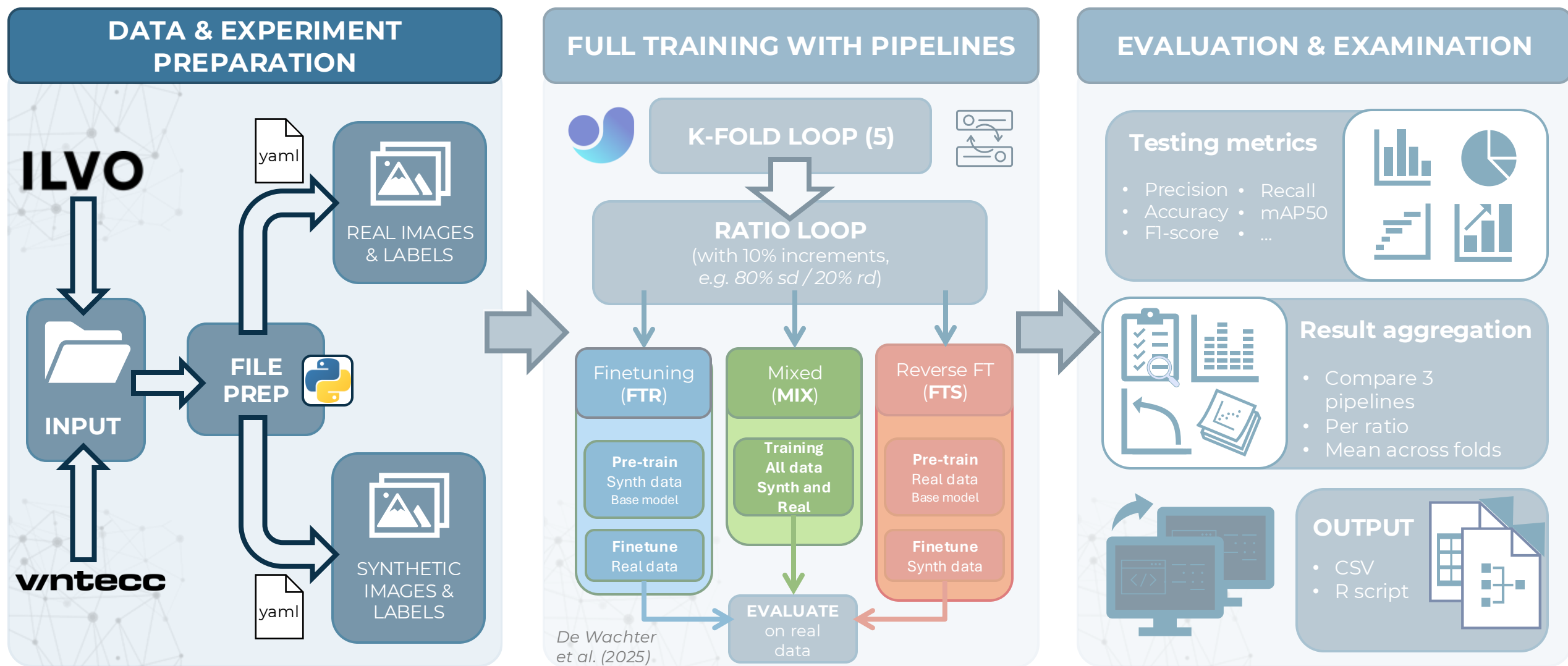
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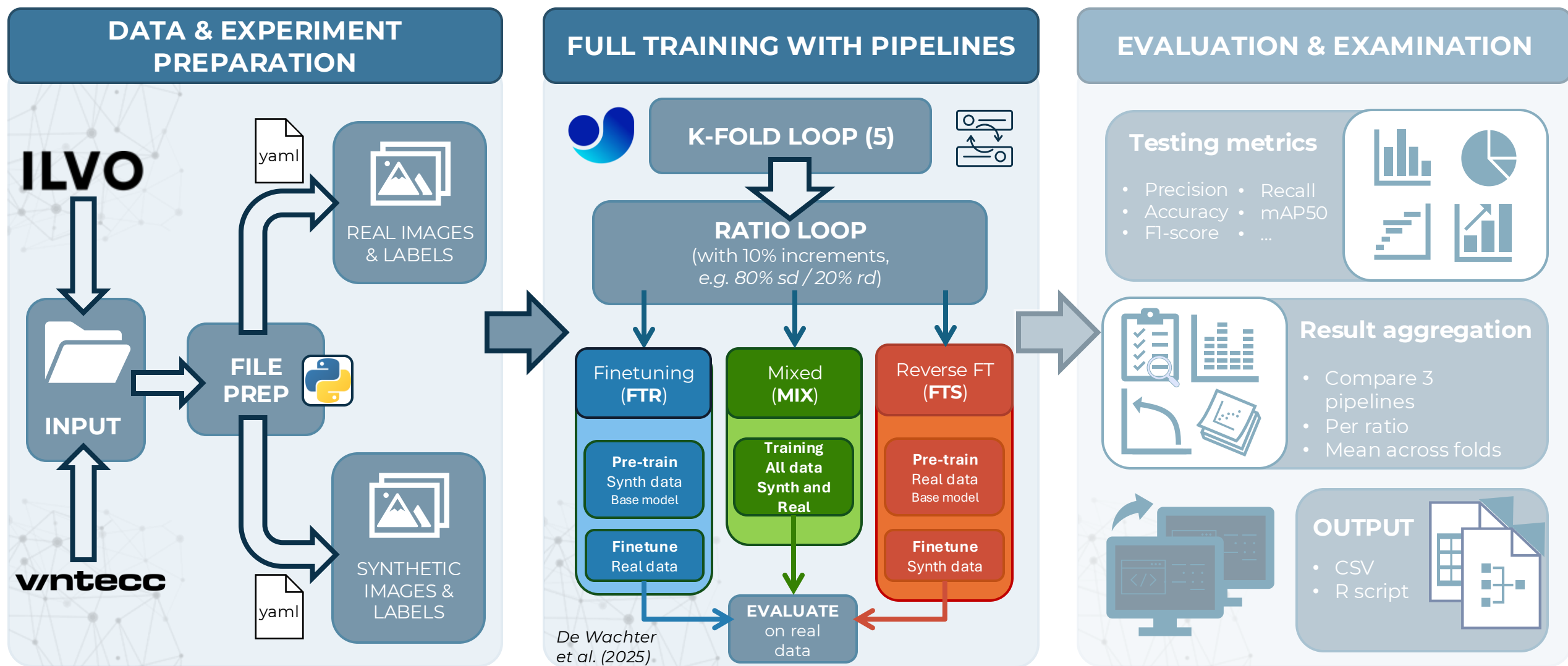
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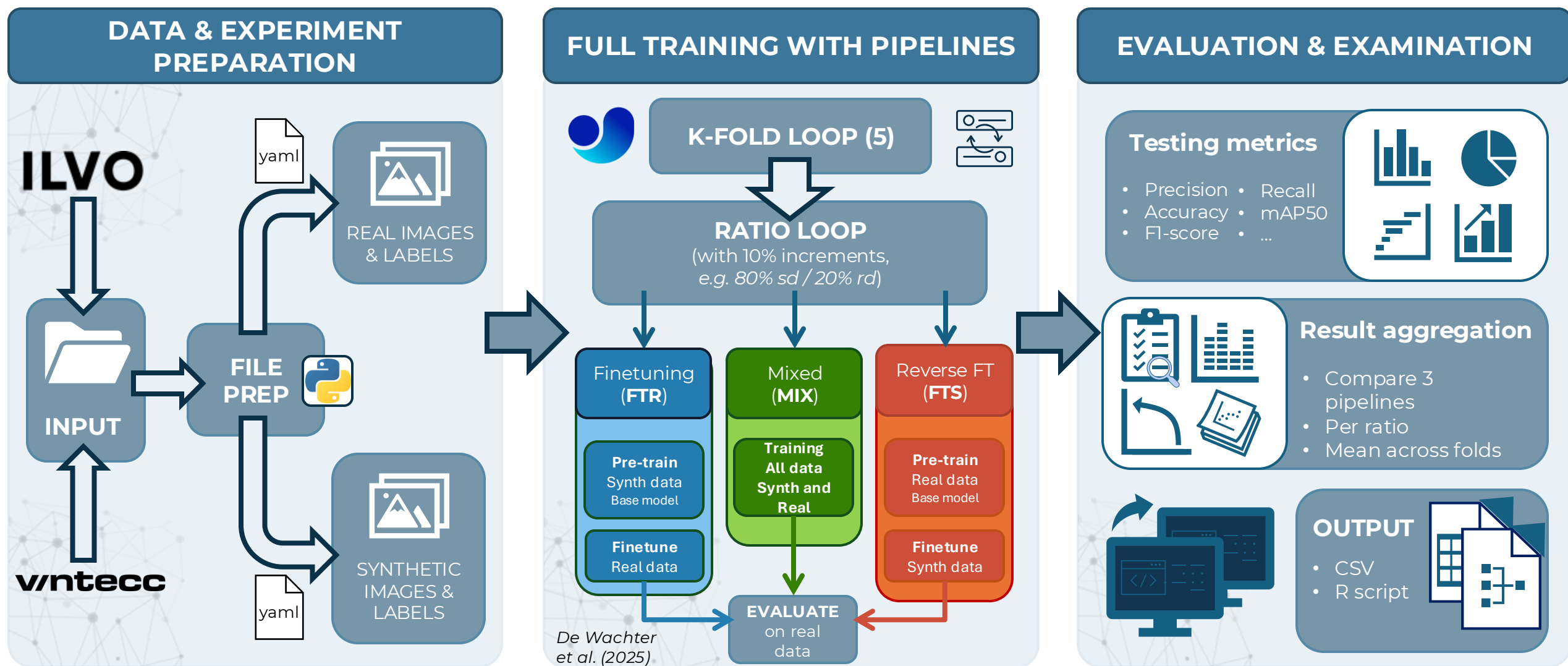
Project workflow

detection & segmentation model training



Project workflow

detection & segmentation model training

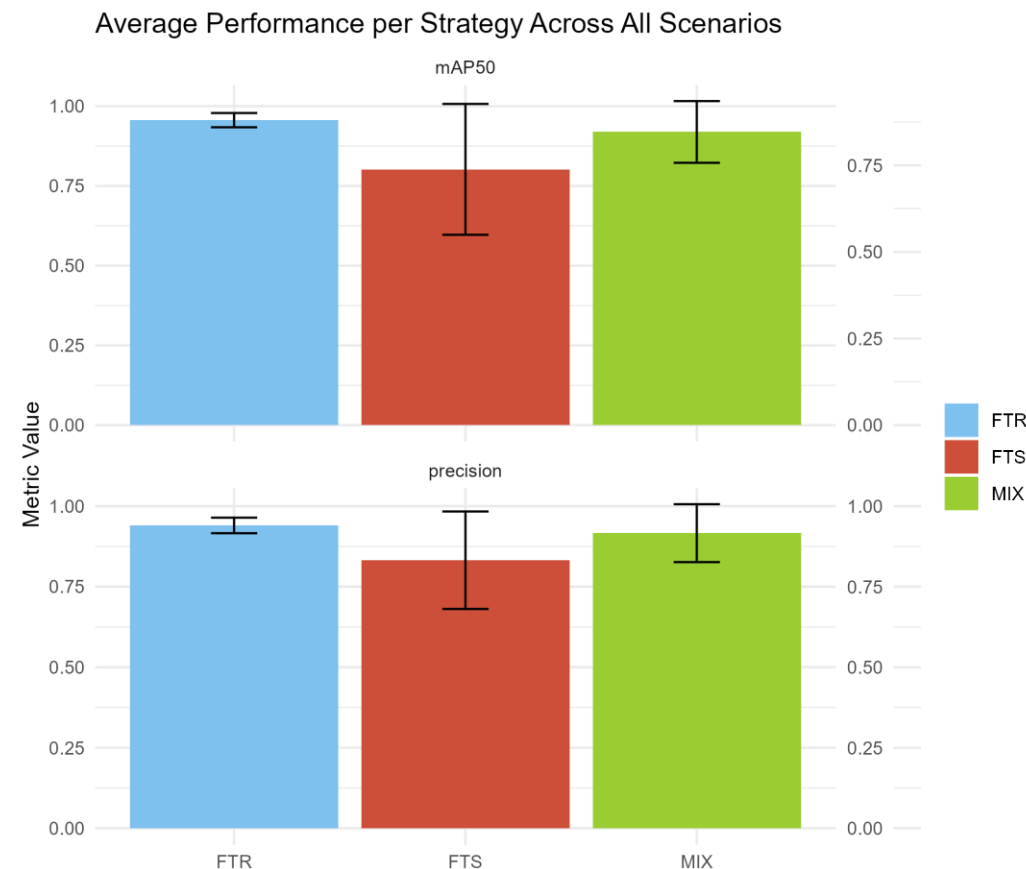


Initial results

YOLO-based

On the **1st synthetic dataset** (*only brill*) results showed:

- Differences between training strategies
- FTR appeared to perform best
- Performance remained high at low–moderate synthetic ratios
- First decline observed around higher synthetic proportions



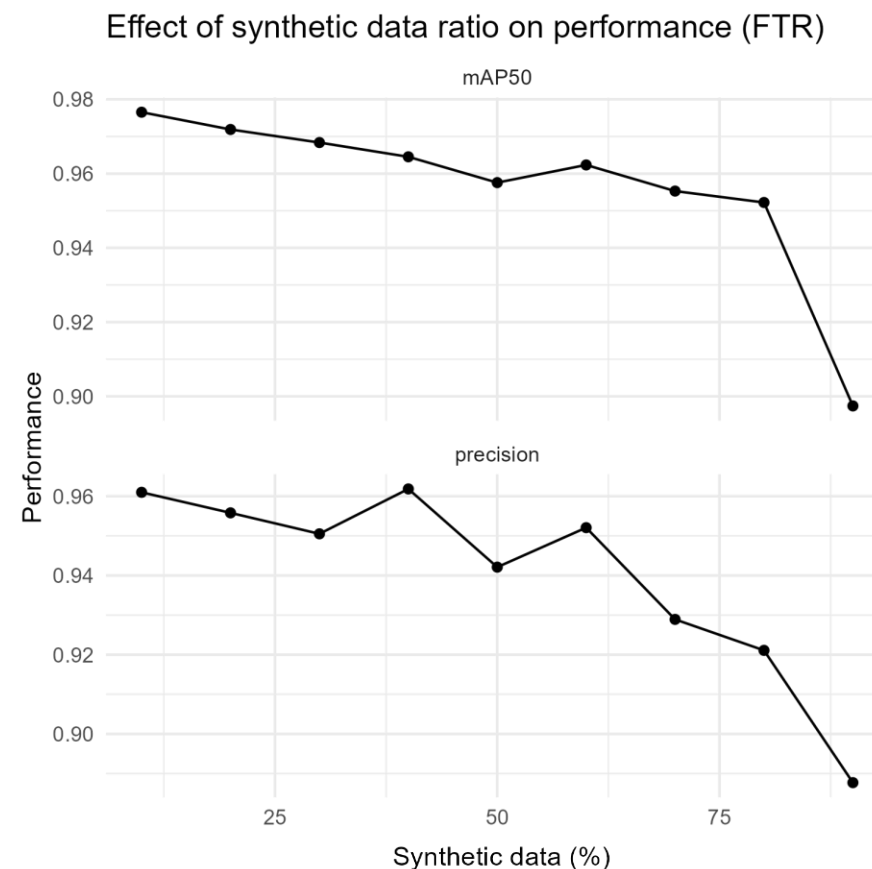
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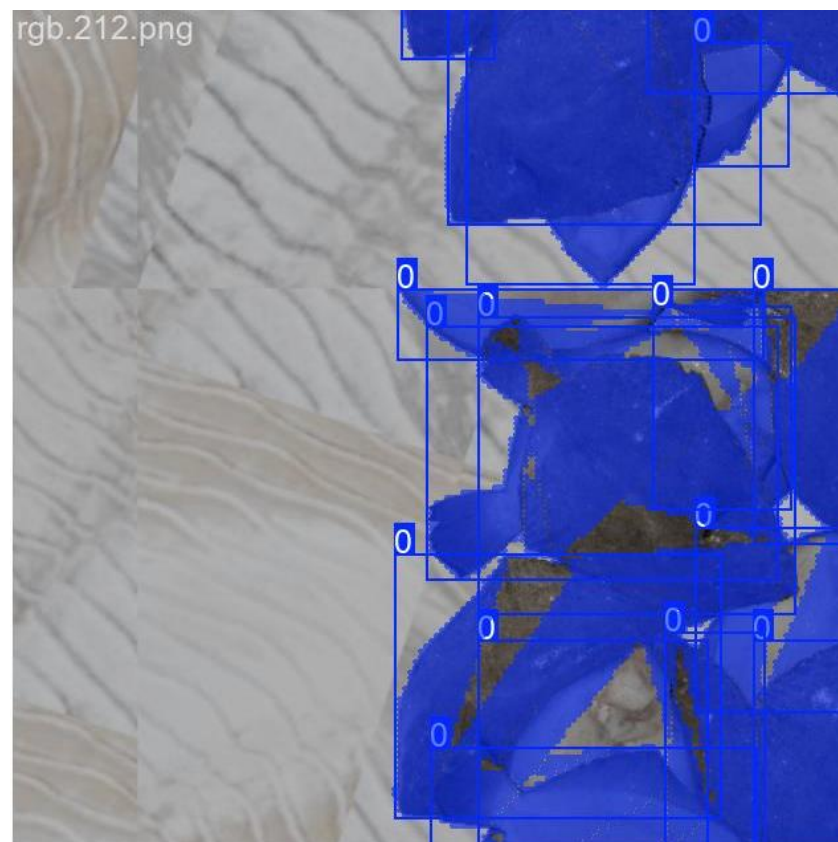
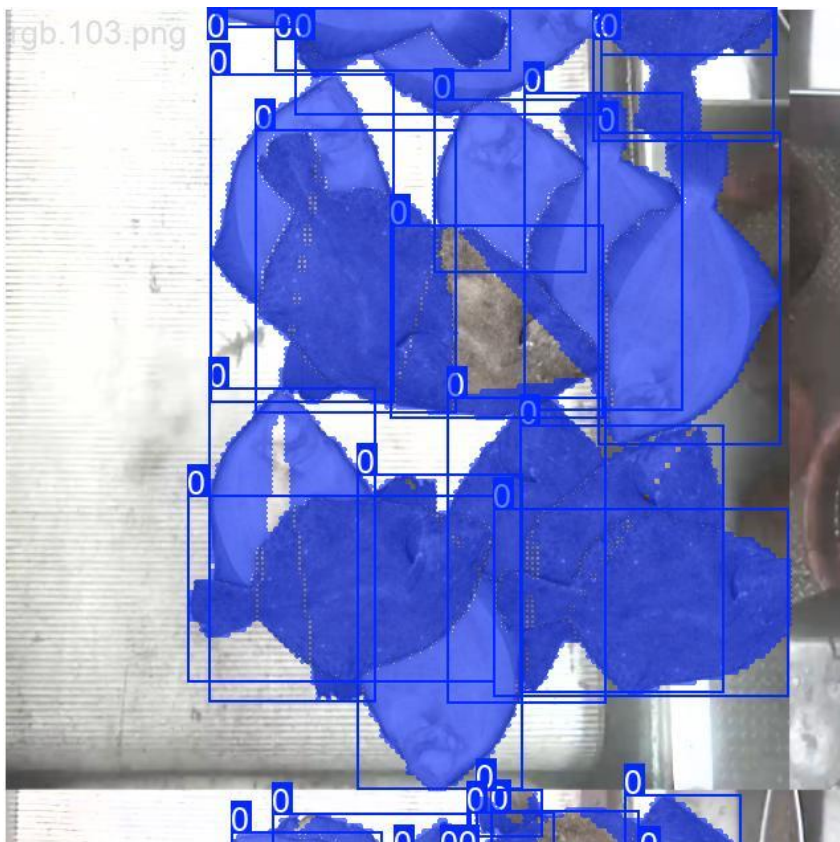
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We still performed a **critical workflow review** before accepting interpretation.



Where workflow sensitivity appeared

Annotation conversion



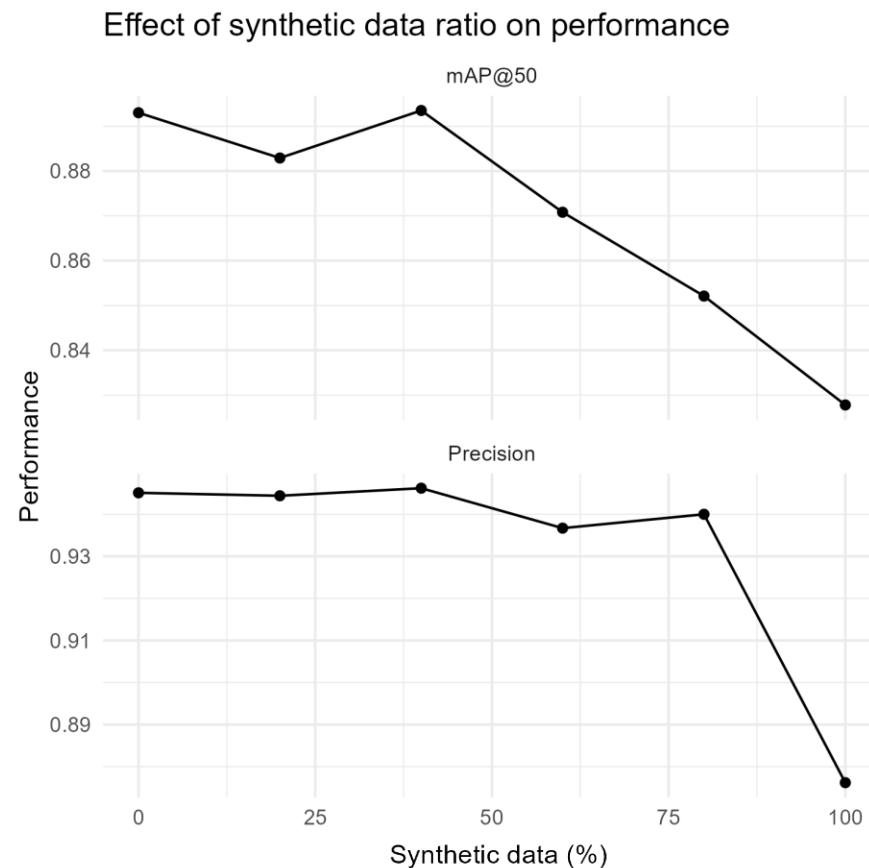
Workflow adjustment

RF-DETR

Current status

- Preliminary analysis
- Mixed (MIX) training only
- Ratios tested at 20% increments

Preliminary trend: performance remains higher up to ~40-60% synthetic data, followed by a decline at higher synthetic proportions.



Next Steps

- ✓ **Re-run all training** branches using the corrected workflow
- ✓ Repeat ratio analysis in **10% increments**
- ✓ **Evaluate robustness** across controlled and operational real-world scenarios
- ✓ **Scale** synthetic data **diversity**:
 - More individuals
 - Plaice as additional species
 - More complex mixed-species scenes
- ✓ Define **where synthetic data adds value**, and where it does not

Take-home Message

- ✓ Synthetic data is **promising**, but **not plug-and-play**
- ✓ Its **value** depends on the **full workflow**: generation, annotation, conversion, training, and validation

SynFish aims to **identify when synthetic data becomes reliable and useful** for AI-based fisheries monitoring, as well as which **workflow choices** determine that **reliability**.

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Thank you for listening!

Any questions?

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