

### A New Application for Automated Video Identification of Marine Species (AVIMS)

**Scottish Government Marine Directorate** 

### Scottish Marine and Freshwater Science Vol 14 No 7

M. Mackiewicz, G. French and M. Fisher

#### Annex 2 – AVIMS Tutorial Slides



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#### A New Application For Automated Video Identification Of Marine Species (AVIMS)

#### Annex 2 – AVIMS Tutorial Slides

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#### Abstract

In the course of its environmental monitoring activities, the Scottish Government's Marine Directorate collects a large amount of underwater video to, for example, obtain information on the numbers of fish in rivers or on species living on the seabed. Manual analysis of this footage is laborious and costly, but Machine Learning algorithms can now be used to automate such image analysis. The Marine Directorate commissioned the University of East Anglia to develop a web-based application to allow staff to create, train and execute machine learning-based (semi-)automated analysis of video footage without a need to interact with the underlying computer code. The application was tested using three diverse sets of video footage and found to be usable by staff without computer science or coding experience. The tool was able to detect and count sea pens in footage from towed underwater vehicles, salmon smolts at sea in underwater footage from towed fishing gear and adult salmon and sea trout in footage from underwater or overhead cameras at fixed locations on rivers. Improving the accuracy of the models at detecting and counting organisms of interest will require the use of larger annotated datasets in further training of the algorithms, but the current application provides a basis for further developing these.

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#### Abbreviations

- AVIMS Automated Video Identification of Marine Species
- CNN Convoluted Neural Network
- GPU Graphics Processing Unit
- GUI Graphical User Interface
- mAP Mean Average Precision
- OS Operating System

## **AVIMS** tutorial

Geoff French University of East Anglia September 2021

## Overview

### Overview

- AVIMS
- Machine learning intro
- Using AVIMS
  - Step 1a: Upload footage
  - Step 1b: Select frames for annotation
  - Step 2: Labelling schema: what objects/creates are we interested in?
  - Step 3: Annotate objects in training images
  - Step 4: Build a training dataset
  - Step 5: Train an object detection model
  - Step 6: Analyse videos
  - Circular work-flow

## AVIMS



• AVIMS is an automated video analysis tool for detecting and counting various creatures or objects of interest in videos

### AVIMS

- Detect then track
  - A creature or object of interest is detected in a video frame
  - If it is detected in a nearby location in subsequent frames it can be tracked
  - Tracking objects across frames supports quantification
  - The objects can be classified, e.g. by species, etc.



### AVIMS

- AVIMS uses machine learning
- This requires the user (you <sup>(i)</sup>) to manually annotate images, identifying objects if interest so that the system can be taught what to look for
- Once trained, the system can be let loose on videos you want analysed

## Machine learning intro

## Machine learning intro

• A machine learning *model* is trained so that given *input* samples, it will predict correct *outputs* 





## Training an ML model

- Construct a dataset consisting of:
  - Input samples
  - Manually prepared ground truths; one for each input
- These are handed to the training software that trains the model to predict the outputs given the inputs
- Then we set the model loose on samples, e.g. images/video that we want analyzed!
  - Well, that's the theory anyway...

## Machine learning – generalization

- Generalization refers to the ability of the model to generate correct predictions given input samples it hasn't seen before
- Machine learning models often *overfit* their training data
  - They often generate very good predictions for samples in the training set
  - Performance on as-of-yet unseen samples is usually lower
  - It would be advantageous to know how well our model will perform on unseen samples...

## Machine learning – train/test split

- We take a dataset of samples with corresponding ground truths and split it into a *training set* and *test set*
- The model is trained using the *training set*
- The samples in the *test set* are held out
- After training, the performance of the model is measured using the *test set*
- In theory, this should give you an idea of how well it will perform in practice...

## Machine learning – train/test split

- In practice, how well *test set* accuracy matches the accuracy obtained in the real world depends on how the *test set* was chosen!
- You need to ensure that the distribution of the samples in the *test set* matches that of the samples the model will be applied to in practice
- If the samples in the *test set* are too similar to those in the *training set* we will get a poor estimate of performance

## Machine learning – train/test split

- How should we split our data? It depends on the use case.
- We will discuss three options
  - Split by site / camera
    - E.g. use site A, B, C, D, E, F for training, G & H for test
  - Split by video
    - E.g. use videos (all from the same camera) 1, 2, 3, 4 for training, 5 & 6 for test
  - Split by image/frame
    - E.g. select frames from within a video and randomly split between train and test

# Machine learning – train/test split – split by camera

### TRAIN





### All images are from different cameras

TEST



# Machine learning – train/test split – split by camera

- Measures the ability of the model to generalize to new sites/viewpoints
  - Probably what is wanted in many cases where you want to conduct surveys at new sites and use a model you trained earlier
- Very challenging from the perspective of the model!
- The training set will need to have images from a number of different cameras/viewpoints in order to have sufficient variety so that the viewpoints in the test set have training samples that are similar enough

# Machine learning – train/test split – split by video

### TRAIN



All images are from the same camera, but from different videos

TEST





# Machine learning – train/test split – split by video

- Less challenging as the model does not have to generalize to new viewpoints
- Sometimes you only have one camera or differentiating between cameras doesn't make sense, e.g. when the camera is moving in an ocean floor survey
- If you can't get enough sites/cameras to get good enough performance, try annotating some frames from videos that are getting poor results

# Machine learning – train/test split – split by frame/image

### TRAIN



All images are from the same video, but from different but close-by frames TEST





# Machine learning – train/test split – split by frame/image

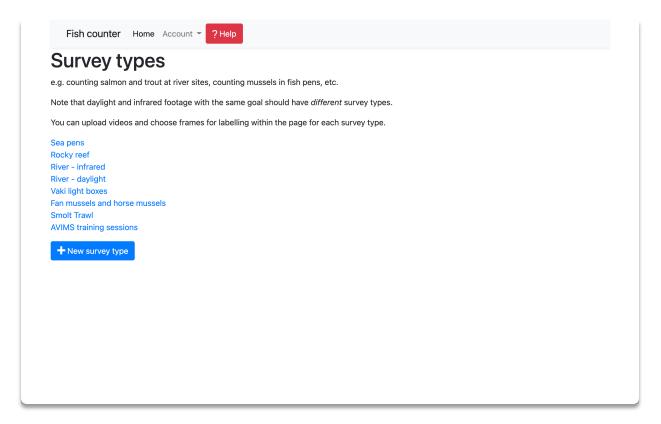
- Least challenging
  - Model performance will be good on close-by or similar frames, but poor on different videos or frames with different content, so watch out!
- If you have 1 long video and you want to annotate say 40 frames and have the model automatically analyse the rest

## Machine learning – cyclic workflow

- In practice, we would:
  - 1. Select samples for our dataset
  - 2. Manually annotate them
  - 3. Split into train/test
  - 4. Train a model
  - 5. Evaluate its performance on novel samples/images/frames/videos/etc.
  - 6. If there are frames on which the model performs unacceptably poorly
    - a. Add them to the dataset
    - b. Go back to step 2 and manually annotate them and repeat the cycle until performance is good enough

## Using AVIMS

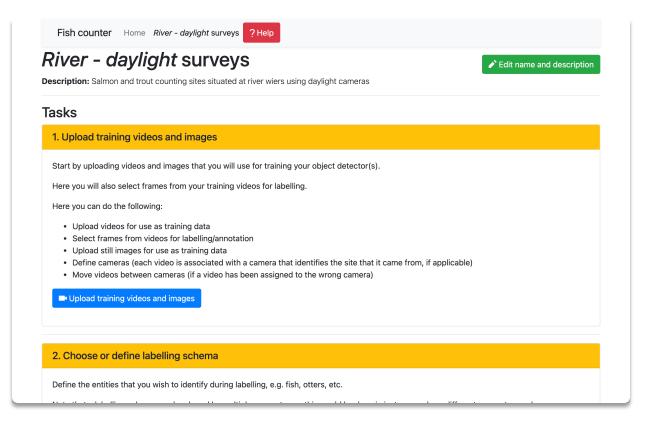
• After logging in you need to choose the survey type you want to work on:



- A survey type:
  - Focuses on detecting a particular range of species or creatures of interest in common conditions/types of location
    - E.g. 'counting salmon and otters at river weir sites' should have a different survey type from the one used to 'count sea pens in underwater footage'
  - Contains footage with a common 'appearance', separating daylight from night-time infrared, underwater from above water
    - E.g. while sharing the common goal of counting salmon and otters at river weir sites, put the daylight and night-time infrared footage in different survey types
    - If we had underwater cameras at river weir sites too, they should go in a different survey type

- When to create a new survey type:
  - Create a new survey type when there isn't one that shares the same goals or the footage has a very different appearance than one of the existing ones.
  - Don't create different survey types for e.g. different years. Split this by different 'cameras' (more on this later)
- Why?
  - All of the data associated with a survey type will be used to train an object detection machine learning (ML) model. Best performance will be obtained by having this model focus on a specific problem and type of *visual appearance*, so group footage by these criteria

• After choosing or creating a survey type you should see this:

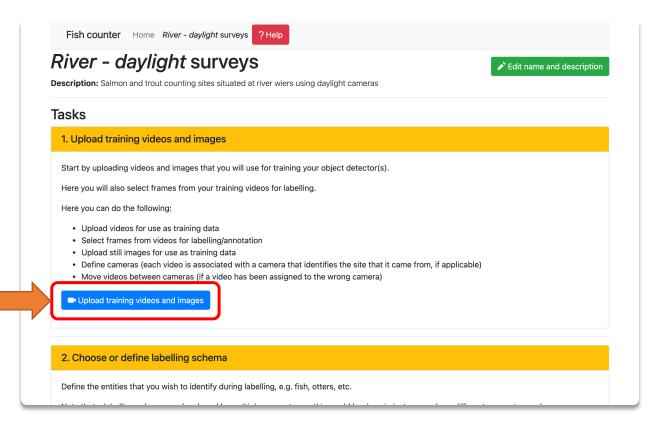


## Using AVIMS

1.a. Upload footage

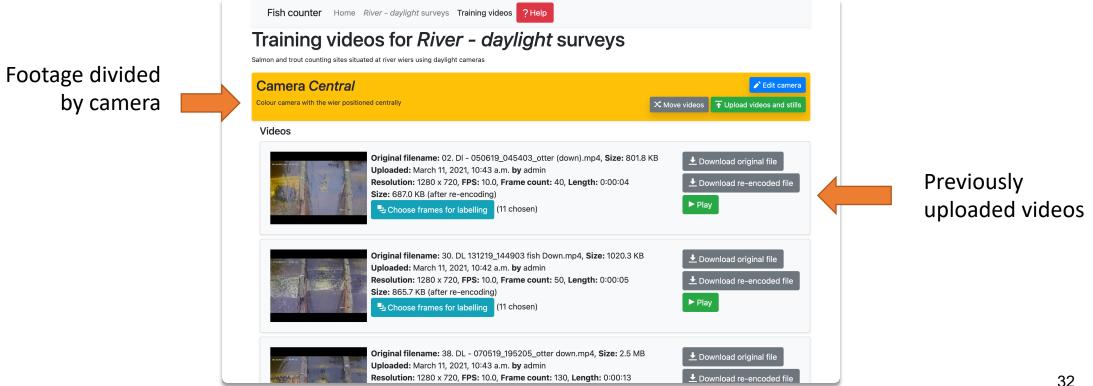
### 1.a. Upload footage

• You need to upload some footage that will be used to train the ML model. From the survey type home, click:



## 1.a. Upload footage

• The training videos page:

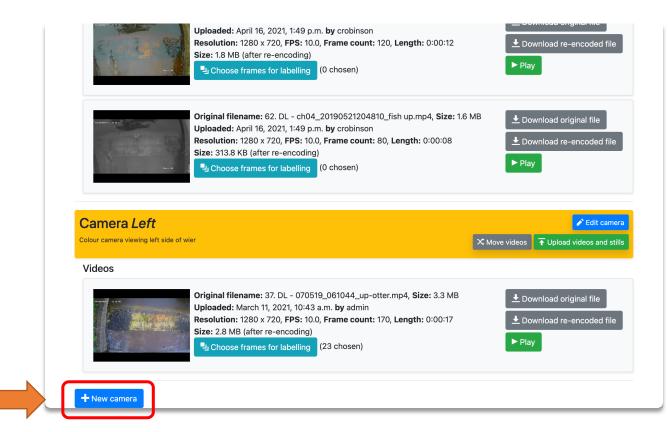


## 1.a. Upload footage

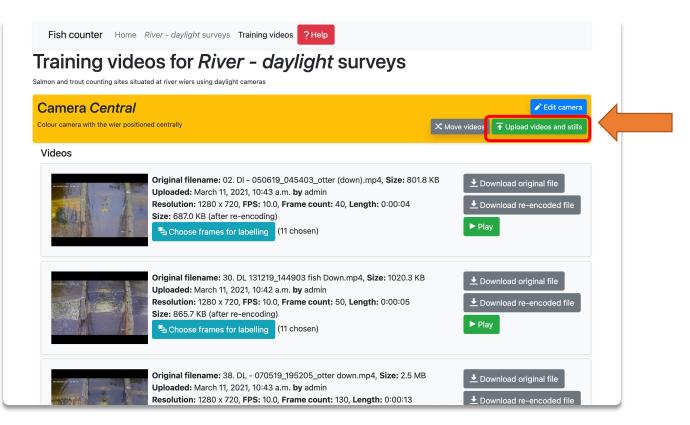
- In the ML intro we mentioned that footage from multiple sites or cameras can be assigned to be in the *training set* or *test set* depending on where it came from
- In this section of the site, you can create a *camera* for each camera/location/site form which footage was obtained

- In seabed surveys conducted using an underwater camera suspended below a vessel, it probably doesn't make sense to divide footage by the *specific device* that was used to record it
- It probably makes more sense to split by location, time of year, or year
- So make a different *camera* for each of these sources of footage.
  - Apologies for the incorrect terminology here; we selected these terms early in the project and they stuck.

• If you need to create a new camera, scroll to the bottom of the upload page and click the *new camera* button:



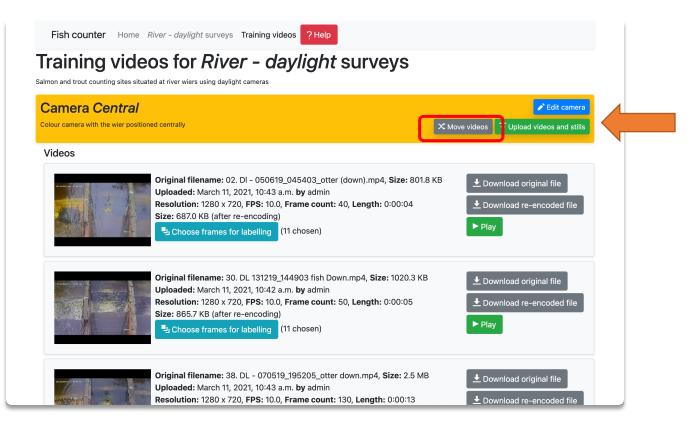
 Choose a camera and click the upload videos and stills button in its' yellow header



 Clicking upload videos and stills button takes you here. You can select files from your computer to upload. Choose either video files, or still images

inages	Fish counter       Home       River - daylight surveys       Training videos       Upload training videos for camera Central         Upload training videos for River - daylight — camera Central         Select files         Please select the files you wish to upload. You can either:	
Click have to open	<ul> <li>Click the blue button and choose files (you can select multiple files)</li> <li>Drag the files from your computer and drop onto the grey box</li> </ul> Please note: it may take a while before large uploads appear in the video list.	
Click here to open a dialog box for locating files on your computer	Choose files to upload Drop files here	Or, drag and drop files from your computer to this area
	Previously uploaded files	
	File name         Upload date         Status           EQ.PM.s64_02_20161007131600_ts.mts         March 18_2021_2:38.p.m         Success: EQ.PM.s64_02_20161007131600_ts.mts	

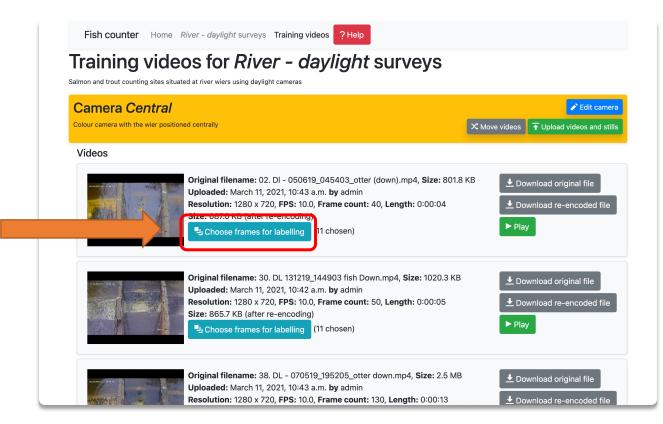
• *NOTE*: if footage has been assigned/uploaded to the wrong camera; the *move videos* button will allow you to change cameras afterwards:



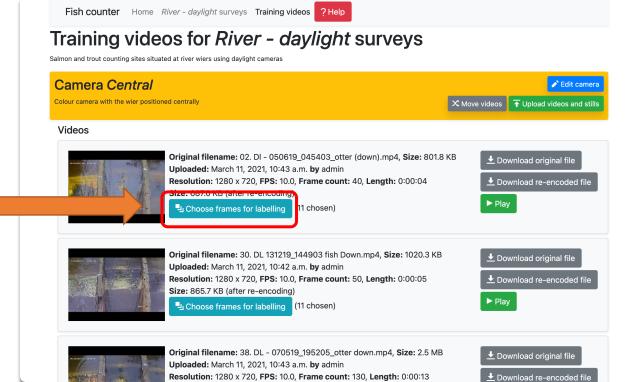
## Using AVIMS

1.b. Select frames for labelling

• From the training videos page, click the *choose frames for labelling* button adjacent to the video you want to select frames from:

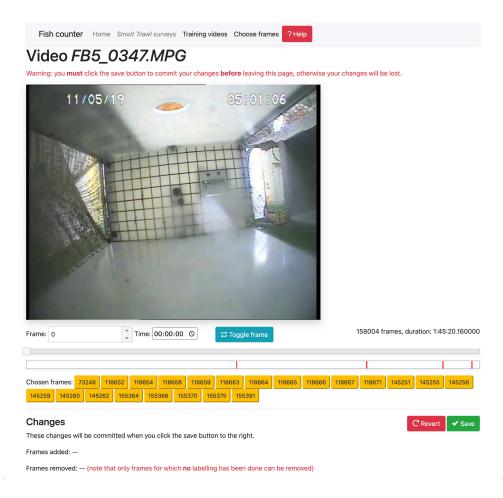


 Note the (11 chosen) to the right of the button; the number of frames chosen from this video. Choose frames from a variety of videos for best results.



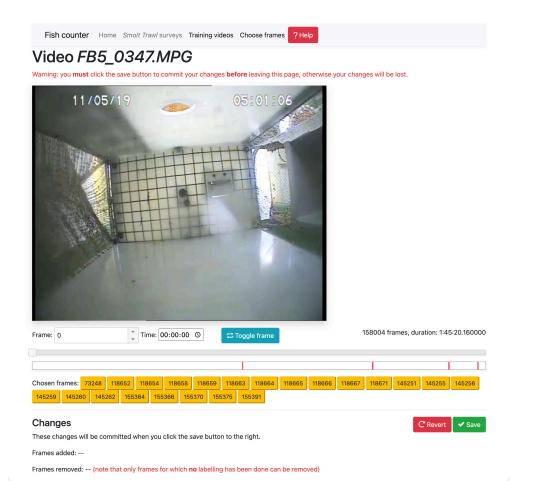
Use the *frame chooser* to choose frames for labelling in step 3.

Pick frames that are as representative as possible of the footage you intend to analyze.



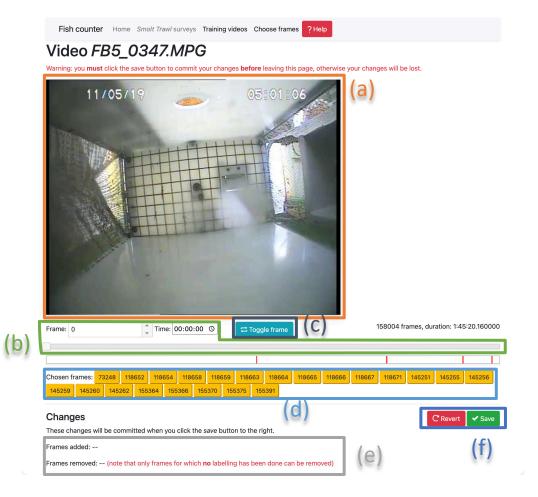
Rules of thumb:

- Choose frames such that as many examples of each kind of creature/objects of interest are visible
- Don't pick too many of the same individual creature/onject
- Try to choose frames in which the appearances (pose, size, angle, colour, etc.) vary as much as possible



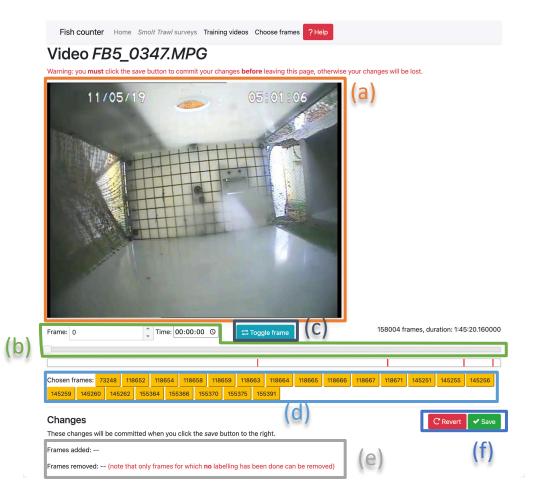
Frame chooser

- a. Image of current frame
- b. Change current frame
- c. Add/remove current frame from the list in (d)
- d. The list of selected frames
- e. The list of changes you have made by selecting/unselecting
- f. Buttons to save or cancel/revert



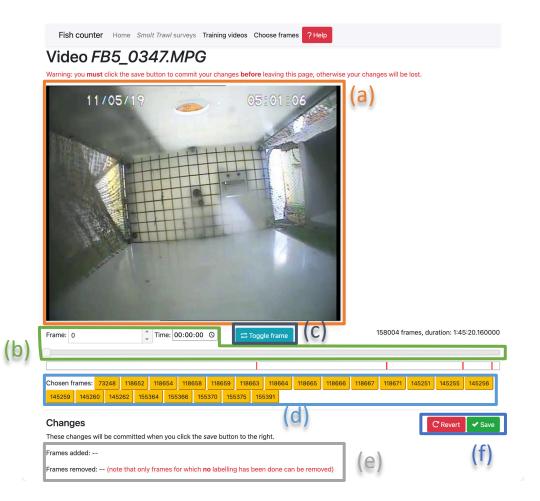
Work-flow:

• NOTE: you must click the *save* button to save your changes once you are done; navigating back to the previous page will result in your changes to the selected frames being **lost**!



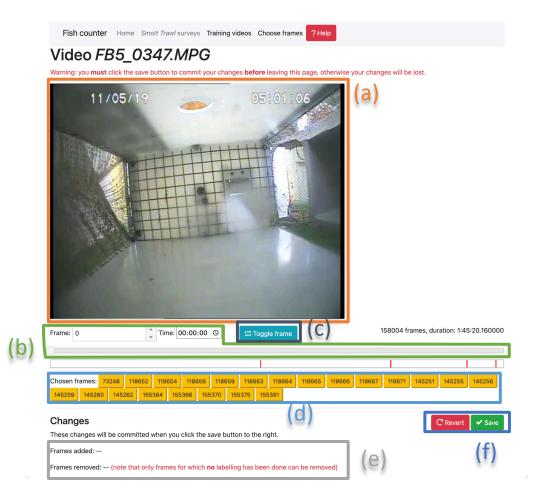
Work-flow:

- Use the slider bar in (b) to sweep through the video and find creatures of interest
  - If the video is long, play it in a video player, find the time at which you see see something you want to annotate then enter the time into the time entry box.
  - Alternatively use the frame entry box



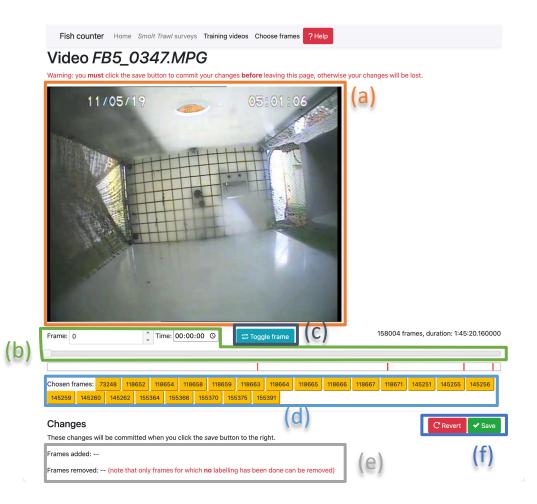
Work-flow:

- Click the toggle frame button (c)
- A red mark will appear in the bar below the slider
- The frame will be added to the list of chosen frames in (d)
- It will also be listed next to *frames added* in (e)



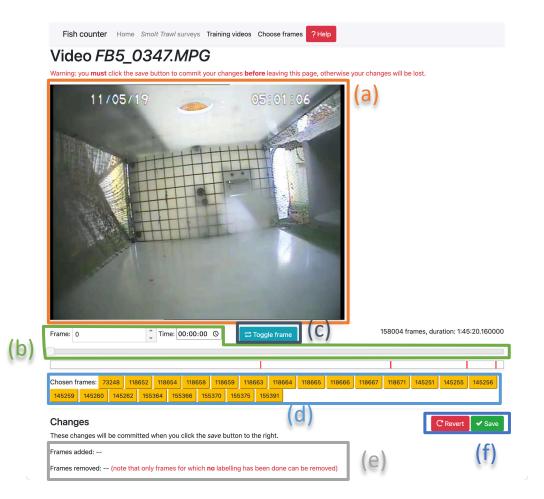
Work-flow:

 Frames that you *unselect* will be added to the list after *frames removed* in (e)



Work-flow:

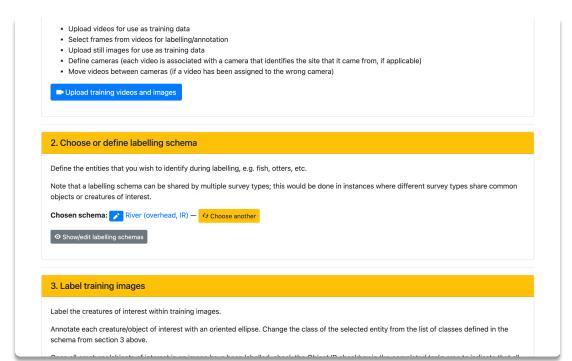
- If you are happy with your changes to the list of selected frames, click the *save* button in (f)
- If you wish to discard your changes, click *revert*



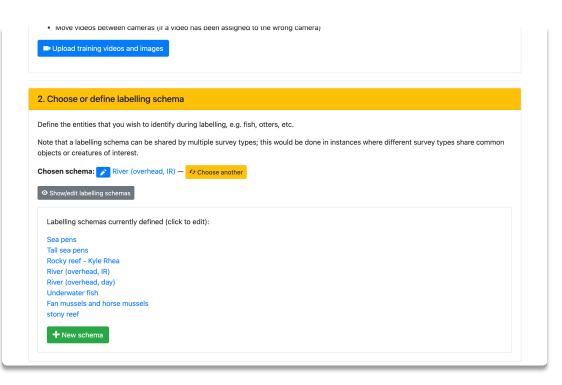
# Using AVIMS

2. Labelling schema: what objects/creates are we interested in?

- Section 2 of the survey page lets you change and edit schemas
- You can edit the schema chosen for the survey type by clicking the blue pen button.
- You can switch schemas by clicking the *choose another* button



- Clicking the show/edit labelling schemas button shows all available schemas, along with the new schema button for creating new ones
- If an existing schema covers the all of creature/object types you are interested in, you can use it
  - Multiple survey types can share a single schema



The schema editor

- The list of types of creature/object defined by the schema are listed as *label classes* in the grey table towards the bottom of the image
- You can add as many label classes as you need to a schema

Fish counter Home Edit label classes in schema Tall sea pens

#### Edit label class in schema Tall sea pens

#### Colour schemes

Different tasks can benefit from using difference colour schemes to help differentiate different labels from one another. For example, a segmentation task could benefit from using different colours for broad categories, while a precise classification/identification task could benefit from using different colours for fine-grained categories within a broad category. Here you can create colour schemes that the users can choose within the labelling tool.

You can re-order the colour schemes by dragging the  $\equiv$  handles.

New	Name	Name in UI
	∃ object_id	Object ID

#### Label classes by group

Label classes are divided into groups. Groups could correspond to broad categories of entity, while individual label classes correspond to more precise classifications. The *default colour* is the default colour that will be used to display the label if no colour schemes are provided. You can assign a colour for each colour scheme defined above.

You can re-order the groups and label classes by dragging the  $\equiv$  handles. Groups can only be deleted if they are empty.

∃ Group:	Tall sea pens			l în centre de la
New	Name	Name in UI	Default colour Object ID	
	≡ tall_sea_pen	Funiculina quadrangula	#f0d10a #808080 💼	
	$\equiv$ fireworks_anomone	Pachycerianthus multip	#eb420a #808080 🟛	
	$\equiv$ celtic_feather_star	Leptometra celtica	#29b158 #808080 🗊	
	$\equiv$ burrowed_mud	Burrowed mud	#cf12d3 #808080 🟛	
New label o	group			

Edit schema name and description

The schema editor

- In the colour schemes section, check if there is an entry in the table with the named object\_id (as seen in the image)
  - If not, click the yellow New button and a form will appear. Enter object\_id for the name field and Object ID for the name in UI field. Then click the green Create button that will have appeared in the form.

Fish counter Home Edit label classes in schema Tall sea pens

#### Edit label class in schema Tall sea pens

#### Colour schemes

Different tasks can benefit from using difference colour schemes to help differentiate different labels from one another. For example, a segmentation task could benefit from using different colours for broad categories, while a precise classification/identification task could benefit from using different colours for fine-grained categories within a broad category. Here you can create colour schemes that the users can choose within the labelling tool.

You can re-order the colour schemes by dragging the  $\equiv$  handles.

New	Name	Name in UI
	$\equiv$ object_id	Object ID

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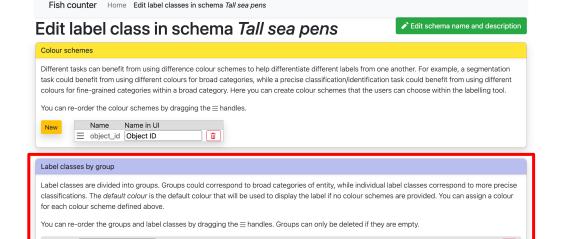
You can re-order the groups and label classes by dragging the  $\equiv$  handles. Groups can only be deleted if they are empty.

lew	Name tall_sea_pen	Name in UI Funiculina quadrangula	Default colour #f0d10a	Object ID #808080	
		Pachycerianthus multip		#808080	
		Leptometra celtica	#29b158	#808080	
	$\equiv$ burrowed_mud	Burrowed mud	#cf12d3	#808080	
/ label	_				

Edit schema name and descriptio

The schema editor

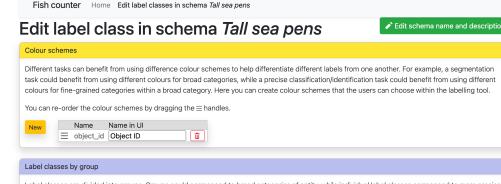
- Mostly you will be adding new label classes and groups in the *label classes by group* section with the lilac header.
  - If you don't see a grey group box (like the one titled Tall sea pens in the image), click the blue New label group button
    - In the form that appears, enter a name and click *create*



	Name tall_sea_pen	Name in UI Funiculina guadrangula	Default colour	Object ID #808080	
		Pachycerianthus multip		#808080	
Ξ	celtic_feather_star	Leptometra celtica	#29b158	#808080	
=	burrowed mud	Burrowed mud	#cf12d3	#808080	

The schema editor

- To add a label class to a group, click the green New button. A form will appear. Fill in the name and name in UI fields and click the create button.
  - You can use these fields how you wish
  - *Name in UI* is what you will see in your list of choices in the labelling tool that we will discuss later



Label classes are divided into groups. Groups could correspond to broad categories of entity, while individual label classes correspond to more precise classifications. The *default colour* is the default colour that will be used to display the label if no colour schemes are provided. You can assign a colour for each colour scheme defined above.

You can re-order the groups and label classes by dragging the  $\equiv$  handles. Groups can only be deleted if they are empty.

=	Name tall_sea_pen	Name in UI Funiculina quadrangula	Delaul	t colour #f0d10a	Object	#808080	<u>ث</u>		
				#eb420a		#808080			
Ξ	celtic_feather_star	Leptometra celtica		#29b158		#808080	<u>ل</u>		
≡	burrowed_mud	Burrowed mud		#cf12d3		#808080	۱.		

#### The schema editor

- The colours in the *default colour* and *Object ID* colour columns in the label classes table define the colours that will be used to distinguish different types of object in the labelling tool and for identifying object types in the videos resulting from analysis
  - Choose colours that are not too similar to one another

Fish counter Home Edit label classes in schema Tall sea pens

#### Edit label class in schema Tall sea pens

#### Colour schemes

Different tasks can benefit from using difference colour schemes to help differentiate different labels from one another. For example, a segmentation task could benefit from using different colours for broad categories, while a precise classification/identification task could benefit from using different colours for fine-grained categories within a broad category. Here you can create colour schemes that the users can choose within the labelling tool.

You can re-order the colour schemes by dragging the  $\equiv$  handles.

New		Name	Name in UI	
	Ξ	object_id	Object ID	<b>1</b>

#### Label classes by group

Label classes are divided into groups. Groups could correspond to broad categories of entity, while individual label classes correspond to more precise classifications. The *default colour* is the default colour that will be used to display the label if no colour schemes are provided. You can assign a colour for each colour scheme defined above.

You can re-order the groups and label classes by dragging the  $\equiv$  handles. Groups can only be deleted if they are empty.

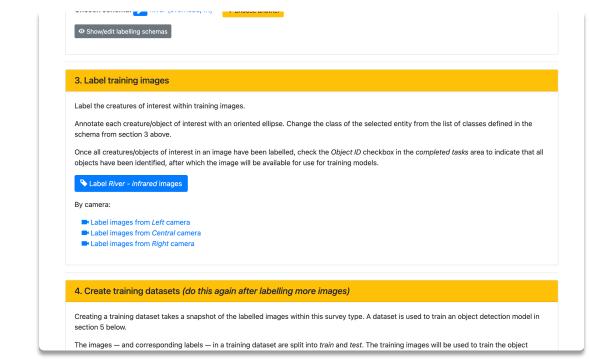
			#f0d10a		#808080	π				
fireworks_anomone	Pachycerianthus multip		#eb420a		#808080	ŵ				
celtic_feather_star	Leptometra celtica		#29b158		#808080	ŵ				
burrowed_mud	Burrowed mud		#cf12d3		#808080	ŵ				
0	celtic_feather_star	celtic_feather_star	celtic_feather_star	celtic_feather_star	celtic_feather_star	celtic_feather_star	celtic_feather_star Leptometra celtica #29b158 #808080 u	celtic_feather_star	celtic_feather_star Leptometra celtica #29b158 #808080 1	celtic_feather_star Leptometra celtica #29b158 #808080 û

Edit schema name and descriptio

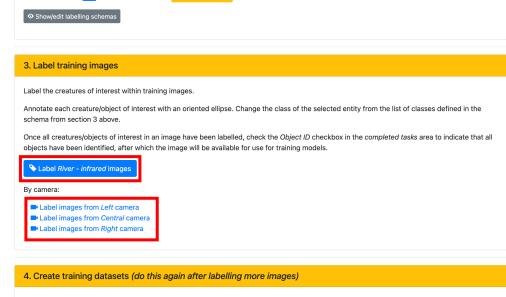
# Using AVIMS

3. Annotate objects in training images

- Section 3 of the survey page has buttons to take you to the annotation part of the site
- Section 3 will be grey and not usable if:
  - You have not uploaded and training videos
  - Or you have not selected any frames for labelling



- You can choose to focus on all training images associated with the survey type (the blue button) or focus on images from each camera (the links below)
  - Clicking either the button or one of the links will take you to...

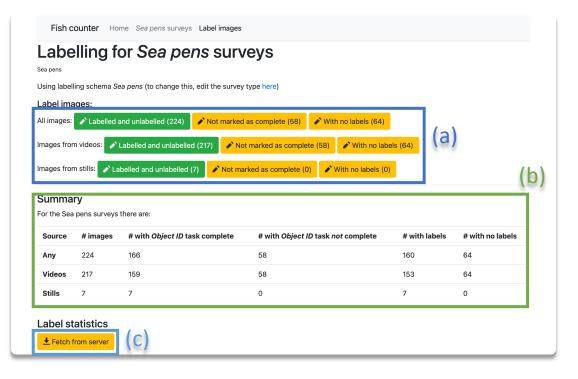


Creating a training dataset takes a snapshot of the labelled images within this survey type. A dataset is used to train an object detection model in section 5 below.

The images – and corresponding labels – in a training dataset are split into *train* and *test*. The training images will be used to train the object

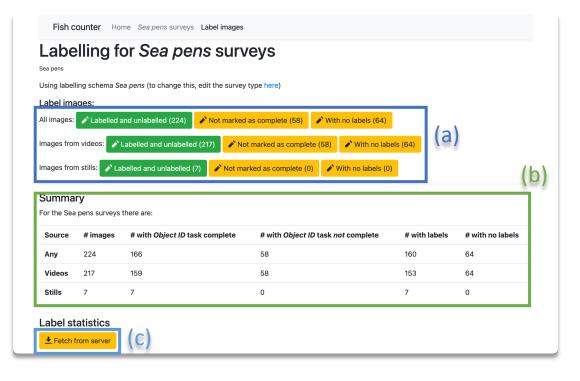
The labelling page

• The buttons in (a) allow you to choose a subset of the images to label. You can choose images for which labelling has not been completed, or images for which no labelling has been done at all, or all images. You can also choose to focus on images from videos or stills.



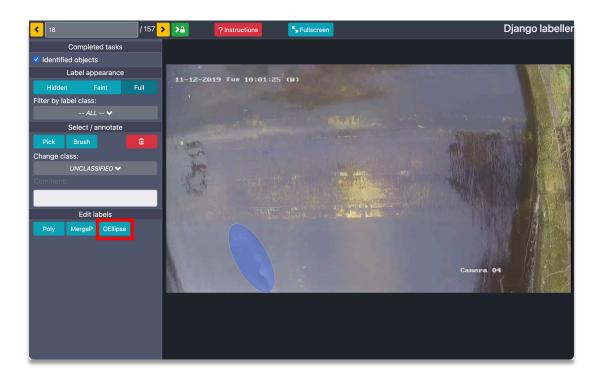
The labelling page

- The summary in (b) gives a brief breakdown of the images
- Pressing the *fetch from server* button (c) under the *label statistics* heading will show you how many of each class of creature/object have been annotated, giving you an idea of which classes are more abundant and which need more instances.



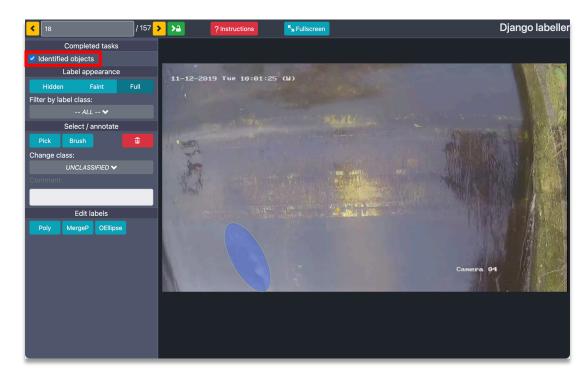
The labelling tool

- The *instructions* button will give you detailed help on how to use the tool
- In general, draw oriented ellipses surrounding objects of interest (OEllipse button)
  - E.g. the ellipse covering the fish in the bottom left of the image
  - Try to get the ellipse to extend from tip to tail along the length of the creature



The labelling tool

- Once all visible objects have ben annotated, click the *identified objects* checkbox to indicate this.
  - If you don't do this, the image will not be used for training; all images not yet marked as *complete* will be ignored



## Using AVIMS

4. Build a training dataset

- Section 4 of the survey page lists the datasets that have been created and when this was done, along with their names and descriptions
- Section 4 will be grey and not usable if:
  - You have not yet labelled any images

#### 4. Create training datasets (do this again after labelling more images)

Creating a training dataset takes a snapshot of the labelled images within this survey type. A dataset is used to train an object detection model in section 5 below.

The images — and corresponding labels — in a training dataset are split into *train* and *test*. The training images will be used to train the object detection model, while the test images will be used to evaluate and quantify its performance.

Typically the samples/images in a training dataset are split between train and test randomly. We give you the option of controlling this more precisely if you wish. You can split by camera, video or by individual image. You can manually assign the individual images or all of the images images from a given camera or video to either train or test.

River infrared - split by image	Likely unrealistically accurate. Not representative of real life usage for short videos like this.	Aug. 15, 2021, 4:01 p.m. <b>by</b> gfrench
River infrared - split by video	Still challenging, representative of conditions in which one or more videos from each site/location can be annotated	Aug. 15, 2021, 4 p.m. <b>by</b> gfrench
River infrared - split by camera	Challenging machine learning conditions representative of deployment at new site or location	June 17, 2021, 5:15 p.m. <b>by</b> gfrench

- Creating a training dataset takes a snapshot of:
  - All of the training images that have been marked as complete
  - Their corresponding labels in the state they are in at the time

#### 4. Create training datasets (do this again after labelling more images)

Creating a training dataset takes a snapshot of the labelled images within this survey type. A dataset is used to train an object detection model in section 5 below.

The images — and corresponding labels — in a training dataset are split into *train* and *test*. The training images will be used to train the object detection model, while the test images will be used to evaluate and quantify its performance.

Typically the samples/images in a training dataset are split between train and test randomly. We give you the option of controlling this more precisely if you wish. You can split by camera, video or by individual image. You can manually assign the individual images or all of the images images from a given camera or video to either train or test.

Description	Created
Likely unrealistically accurate. Not representative of real life usage for short videos like this.	Aug. 15, 2021, 4:01 p.m. <b>by</b> gfrench
Still challenging, representative of conditions in which one or more videos from each site/location can be annotated	Aug. 15, 2021, 4 p.m. <b>by</b> gfrench
Challenging machine learning conditions representative of deployment at new site or location	June 17, 2021, 5:15 p.m. <b>by</b> gfrench
	Likely unrealistically accurate. Not representative of real life usage for short videos like this. Still challenging, representative of conditions in which one or more videos from each site/location can be annotated Challenging machine learning conditions representative

#### • If you:

- Select more video frames for labelling
- Label more images
- Change the labels in the images currently marked as complete (e.g. to fix mistakes)
- You will need to create a *new* training dataset to take advantage of them when training a model

#### 4. Create training datasets (do this again after labelling more images)

Creating a training dataset takes a snapshot of the labelled images within this survey type. A dataset is used to train an object detection model in section 5 below.

The images – and corresponding labels – in a training dataset are split into *train* and *test*. The training images will be used to train the object detection model, while the test images will be used to evaluate and quantify its performance.

Typically the samples/images in a training dataset are split between train and test randomly. We give you the option of controlling this more precisely if you wish. You can split by camera, video or by individual image. You can manually assign the individual images or all of the images images from a given camera or video to either train or test.

River infrared - split by image	Likely unrealistically accurate. Not representative of real life usage for short videos like this.	Aug. 15, 2021, 4:01 p.m. <b>by</b> gfrench
River infrared - split by video	Still challenging, representative of conditions in which one or more videos from each site/location can be annotated	Aug. 15, 2021, 4 p.m. <b>by</b> gfrench
River infrared - split by camera	Challenging machine learning conditions representative of deployment at new site or location	June 17, 2021, 5:15 p.m. <b>by</b> gfrench

- Clicking one of the blue links will allow you to edit the name and description of the dataset
  - You will not be able to alter its contents as it is a fixed snapshot
- Create a new dataset by clicking the green Create a new dataset button

#### 4. Create training datasets (do this again after labelling more images)

Creating a training dataset takes a snapshot of the labelled images within this survey type. A dataset is used to train an object detection model in section 5 below.

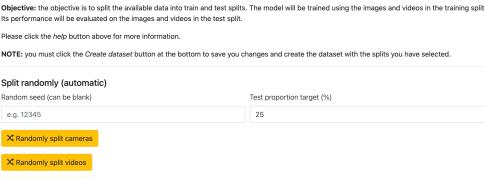
The images — and corresponding labels — in a training dataset are split into *train* and *test*. The training images will be used to train the object detection model, while the test images will be used to evaluate and quantify its performance.

Typically the samples/images in a training dataset are split between train and test randomly. We give you the option of controlling this more precisely if you wish. You can split by camera, video or by individual image. You can manually assign the individual images or all of the images images from a given camera or video to either train or test.

River infrared - split by image	Likely unrealistically accurate. Not representative of real life usage for short videos like this.	Aug. 15, 2021, 4:01 p.m. <b>by</b> gfrench
River infrared - split by video	Still challenging, representative of conditions in which one or more videos from each site/location can be annotated	Aug. 15, 2021, 4 p.m. <b>by</b> gfrench
River infrared - split by camera	Challenging machine learning conditions representative of deployment at new site or location	June 17, 2021, 5:15 p.m. <b>by</b> gfrench

Dataset builder (random; top)

- The dataset builder lets you split the data into *train* and test (see discussion in the ML intro)
- The fastest way of setting the splits is to have the machine split the data randomly



? Help

Fish counter Home River - infrared surveys Create dataset

Create dataset for River - infrared surveys

Note: cameras or videos that have a grey lock 🔒 symbol besides them below will not be affected / randomly assigned by clicking these buttons.

Manually edit splits		
Cameras Videos Images	# imgs	Split
🗖 Left	22	Train Test 🔀
■ 11. IR - ch01_20191112022303 Down fish .mp4	15	Train Test 🔀

Dataset builder (random; top)

- The buttons in (b) will randomly assign the split by camera, video or individual image
  - E.g. clicking *Randomly split cameras* will assign the same split (train or test) to all of the images from videos from that camera
  - Similarly for Randomly split videos

Fish counter Home River - infrared surveys Create dataset ?Help

#### Create dataset for River - infrared surveys

**Objective:** the objective is to split the available data into train and test splits. The model will be trained using the images and videos in the training split. Its performance will be evaluated on the images and videos in the test split.

Please click the *help* button above for more information.

NOTE: you must click the Create dataset button at the bottom to save you changes and create the dataset with the splits you have selected.

#### Split randomly (automatic) Random seed (can be blank) Test proportion target (%) e.g. 12345 25 (a) X Randomly split cameras X Randomly split videos 🗙 Randomly split imag Note: cameras or videos that have a grey lock 🔒 symbol besides them below will not be affected / randomly assigned by clicking these buttons. Manually edit splits Cameras Videos Images # imas Split Train Test 🗙 🖸 Left 22 X ■ 11. IR - ch01\_20191112022303 Down fish .mp4 15 Train Test

Dataset builder (random; top)

- When splitting randomly
  - You can determine the proportion of the data that will be assigned to the *test* split by changing the *Test* proportion target field in (a)
  - If the Random seed field (in (a)) is blank, a different split will be generated each time. Giving a value will control the random split that is used.

Fish counter Home River - infrared surveys Create dataset ? Help

#### Create dataset for River - infrared surveys

**Objective:** the objective is to split the available data into train and test splits. The model will be trained using the images and videos in the training split. Its performance will be evaluated on the images and videos in the test split.

Please click the *help* button above for more information.

NOTE: you must click the Create dataset button at the bottom to save you changes and create the dataset with the splits you have selected.

#### Split randomly (automatic) Random seed (can be blank) Test proportion target (%) e.g. 12345 25 (a) X Randomly split cameras X Randomly split videos 🗙 Randomly split imag Note: cameras or videos that have a grey lock 🔒 symbol besides them below will not be affected / randomly assigned by clicking these buttons. Manually edit splits Cameras Videos Images # imas Split Train Test 🗙 🖸 Left 22 ■ 11. IR - ch01\_20191112022303 Down fish .mp4 15 Train Test X

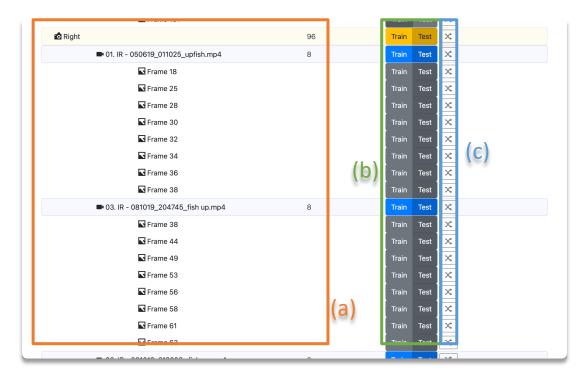
- Further down you will find manual controls for the splits for each image in the dataset
- The images in (a) are divided by camera and video

n Right	96		Train	Test	$\times$
■ 01. IR - 050619_011025_upfish.mp4	8		Train	Test	×
Frame 18			Train	Test	×
Frame 25			Train	Test	×
Frame 28			Train	Test	×
Frame 30			Train	Test	×
Frame 32			Train	Test	×
Frame 34			Train	Test	× (C)
Frame 36		(b)	Train	Test	×
Frame 38		1 -	Train	Test	×
• 03. IR - 081019_204745_fish up.mp4	8		Train	Test	×
Frame 38			Train	Test	×
Frame 44			Train	Test	×
Frame 49			Train	Test	×
Frame 53			Train	Test	×
Frame 56			Train	Test	×
Frame 58		(a)	Train	Test	×
Frame 61			Train	Test	×
Eromo 62			Train	Toot	~*

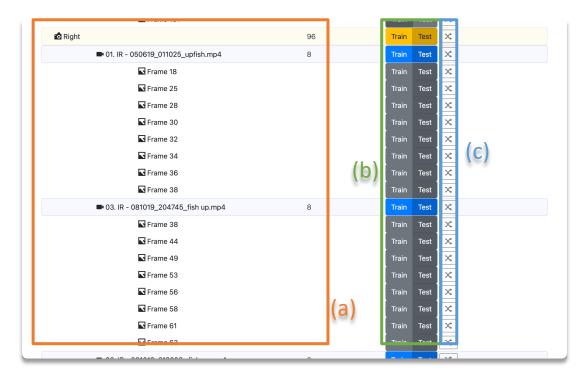
- You can assign images to train or test by clicking the *Train or Test* buttons in (b).
  - Clicking the grey buttons will assign the split of an individual image.
  - Clicking the blue or yellow buttons will assign the split to all images in the given video or camera respectively

🖸 Right	96		Train	Test	$\times$	
01. IR - 050619_011025_upfish.mp4	8		Train	Test	$\times$	
🖬 Frame 18			Train	Test	$\times$	
🖬 Frame 25			Train	Test	×	
🖬 Frame 28			Train	Test	×	
🖬 Frame 30			Train	Test	$\times$	
Frame 32			Train	Test	$\times$	
Frame 34			Train	Test	$\times$	(C)
Frame 36		(b)	Train	Test	×	1-1
Frame 38		5-5	Train	Test	$\times$	
• 03. IR - 081019_204745_fish up.mp4	8		Train	Test	$\times$	
Frame 38			Train	Test	×	
Frame 44			Train	Test	×	
Frame 49			Train	Test	$\times$	
Frame 53			Train	Test	$\times$	
Frame 56			Train	Test	×	
Frame 58		(a)	Train	Test	$\times$	
Frame 61			Train	Test	$\times$	
E France 62			Train	Toot	~*	

- Clicking the buttons in (c) will change the button to a lock icon. Locked images/videos/cameras will be unaffected if you click the split randomly buttons further up, as discussed previously
  - This allows you to specifically assign a split to certain images/videos/cameras, while allowing the rest to be determined randmoly



- Clicking the buttons in (c) will change the button to a lock icon. Locked images/videos/cameras will be unaffected if you click the split randomly buttons further up, as discussed previously
  - This allows you to specifically assign a split to certain images/videos/cameras, while allowing the rest to be determined randmoly



Dataset builder

- Finally, give the dataset a name and a description
- Click the green *Create image dataset* button to create it
  - As stated earlier, once created, only its name and description can be edited; the splits assigned to the images *cannot* be modified later

Frame 170	Train Test 🔀
Frame 172	Train Test 🔀
Frame 173	Train Test 🔀
■ Frame 174	Train Test 🔀
■ Frame 175	Train Test 🔀
Image: Second Secon	Train Test 🔀
95 training images, 96 test images (test proportion: 50.3%)	
Create dataset	
Name	
Description	
+ Create image dataset Cancel	

# Using AVIMS

5. Train an object detection model

- Section 5 of the survey page lists the models that have been trained.
- Section 5 will be grey and not usable if:
  - You have not created any datasets

#### 5. Train an object detection model

Using one of the training datasets listed in section 4 above, train an object detection model that will be used to analyse videos and generate results. Click the *train an object detection model* button and choose a training dataset to use. You will be taken to a page that allows you to modify the settings used to train the model.

Training a model often takes 1 hour or more. If other training or analysis jobs have been queued, your model will wait in the queue until other jobs have been completed.

While a model is training and after training is complete you may click its *report* button to view graphs showing the how the accuracy of the model progressed during training. It also contains logs that can be used by the development team to assess problems.

Once training is complete, the view predictions button will allow you to view the results of applying the model to the individual frames and still images that are in the test set. Clicking this will show the test set images in the annotation tool, with annotations automatically generated using the model. This will give you an idea of the kind of objects that the model is successfully detecting or missing. You may notice the detection failures correlate with certain appearance, e.g. fish with a curved pose. Under the *analysis results* column you will see a link that will take you to the results of all analyses that have been performed using this model.

River infrared	River infrared - split by image	Ready for use	mAP=55.685% • View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	0 analyses
River infrared - split by video	River infrared - split by video	Ready for use	MAP=51.428% View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	4 analyses
River infrared	River infrared - split by camera	Ready for use	mAP=45.556% • View predictions	Aug. 17, 2021, 3:54 a.m. <b>by</b> gfrench	0 analyses

- Each model in the list shows:
  - The name of the model, with an edit button to change its' name
  - The dataset used to train it
  - A training report and log
  - Accuracy and predictions on the test set
  - Links to any analysis results generated using the model (see section 6, later)

#### 5. Train an object detection model

Using one of the training datasets listed in section 4 above, train an object detection model that will be used to analyse videos and generate results. Click the *train an object detection model* button and choose a training dataset to use. You will be taken to a page that allows you to modify the settings used to train the model.

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River infrared	River infrared - split by image	Ready for use	mAP=55.685% • View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	0 analyses
River infrared - split by video	River infrared - split by video	Ready for use	mAP=51.428% © View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	4 analyses
River infrared - split by camera	River infrared - split by camera	Ready for use	mAP=45.556% • View predictions	Aug. 17, 2021, 3:54 a.m. <b>by</b> gfrench	0 analyses

- To train an object detection model, click the green *Train an object detection model...* button
  - A menu will appear allowing you to select the dataset you wish to use
  - Choose a dataset

Once training is complete, the view predictions button will allow you to view the results of applying the model to the individual frames and still images that are in the test set. Clicking this will show the test set images in the annotation tool, with annotations automatically generated using the model. This will give you an idea of the kind of objects that the model is successfully detecting or missing. You may notice the detection failures correlate with certain appearance, e.g. fish with a curved pose. Under the *analysis results* column you will see a link that will take you to the results of all analyses that have been performed using this model.

Name	Dataset	Status	Accuracy	Created	Analysis results
River infrared - split by image	River infrared - split by image	Ready for use	mAP=55.685% View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	0 analyses
River infrared - split by video	River infrared - split by video	Ready for use	mAP=51.428% © View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	4 analyses
<ul> <li>River infrared</li> <li>split by camera</li> </ul>	River infrared - split by camera	Ready for use	mAP=45.556% • View predictions	Aug. 17, 2021, 3:54 a.m. <b>by</b> gfrench	0 analyses
using River infrared	tion model  - split by image dataset - split by video dataset - split by camera datase				
e one of the object det	ection models listed in s	ection 5 above to ana	lyse videos and generate	reports.	
	os button will take you to	the second state of the se			

#### • The train model form:

- Fill in the name and description
- The model type allows you to choose between detecting objects as oriented ellipses, or nonoriented boxes
- The default settings for *Epochs* and *Score threshold* should work in most cases. You can try other settings if you wish

HISH COUNTER Home surveys Train a model for dataset River infrared - split by image						
Train a model for datas	set <i>River in</i>	frared - split by image				
Likely unrealistically accurate. Not representative of real life usa	ge for short videos like this.					
Dataset size: 143 training samples, 48 test samples						
Name						
Description						
		b				
Model type		Ellipse (Mask R-CNN + simplify)				
Mask R-CNN can detect precise outlines that we simplify to ellips can only detected rectangles	ses, while Faster R-CNN					
Epochs		100				
Train for this many epochs. After each epoch the loss and model reported in the log.	performance will be					
reported in the log.						
Score threshold		0.5				
Inference: the detection threshold used during inference, betwee prediction score is less than this value will; be ignored as false de						
Show advanced training settings						
🖁 Train 🗶 Cancel						

#### • The train model form:

- The yellow Show advanced training settings button will show more fields. These can be left as is.
- Speak to the developers about using different settings here
- Click the green *Train* button to train the model
  - Training takes time and will happen in the background

	River infrared - split by image
Likely unrealistically accurate. Not representative of real life usage for short v	videos like this.
Dataset size: 143 training samples, 48 test samples	
Name	
Description	
Model type	Ellipse (Mask R-CNN + simplify)
Mask R-CNN can detect precise outlines that we simplify to ellipses, while Fas	ster R-CNN
Mask R-CNN can detect precise outlines that we simplify to ellipses, while Fas can only detected rectangles	iter R-CNN
	100
can only detected rectangles	100
can only detected rectangles Epochs Train for this many epochs. After each epoch the loss and model performance	100
can only detected rectangles Epochs Train for this many epochs. After each epoch the loss and model performance	100
can only detected rectangles Epochs Train for this many epochs. After each epoch the loss and model performance reported in the log.	2 will be
can only detected rectangles Epochs Train for this many epochs. After each epoch the loss and model performance reported in the log. Score threshold Inference: the detection threshold used during inference, between 0 and 1. Ot	2 will be
can only detected rectangles Epochs Train for this many epochs. After each epoch the loss and model performance reported in the log. Score threshold Inference: the detection threshold used during inference, between 0 and 1. Ot	2 will be
can only detected rectangles Epochs Train for this many epochs. After each epoch the loss and model performance reported in the log. Score threshold Inference: the detection threshold used during inference, between 0 and 1. Of prediction score is less than this value will; be ignored as false detections.	2 will be

 The status will start out as Queued, indicating that the training job has not started.
 Once it starts, the yellow Report button will appear. You can click this during training to bring up a progress graph

#### 5. Train an object detection model

Using one of the training datasets listed in section 4 above, train an object detection model that will be used to analyse videos and generate results. Click the *train an object detection model* button and choose a training dataset to use. You will be taken to a page that allows you to modify the settings used to train the model.

Training a model often takes 1 hour or more. If other training or analysis jobs have been queued, your model will wait in the queue until other jobs have been completed.

While a model is training and after training is complete you may click its *report* button to view graphs showing the how the accuracy of the model progressed during training. It also contains logs that can be used by the development team to assess problems.

Once training is complete, the view predictions button will allow you to view the results of applying the model to the individual frames and still images that are in the test set. Clicking this will show the test set images in the annotation tool, with annotations automatically generated using the model. This will give you an idea of the kind of objects that the model is successfully detecting or missing. You may notice the detection failures correlate with certain appearance, e.g. fish with a curved pose. Under the *analysis results* column you will see a link that will take you to the results of all analyses that have been performed using this model.

River infrared - split by image	River infrared - split by image	Ready for use	mAP=55.685% • View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	0 analyses
River infrared - split by video	River infrared - split by video	Ready for use	MAP=51.428% © View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	4 analyses
River infrared	River infrared - split by camera	Ready for use	mAP=45.556% • View predictions	Aug. 17, 2021, 3:54 a.m. <b>by</b> gfrench	0 analyses

- Once training has finished, the status will change to *Ready for Use*, as seen in the image to the right
  - Under the accuracy column you will see a measure of the accuracy as *mean average precision* (mAP)
  - The View predictions button will be available. Clicking it will show the images in the dataset test split in the labelling tool, with the labels that were automatically predicted by the model

#### 5. Train an object detection model

Using one of the training datasets listed in section 4 above, train an object detection model that will be used to analyse videos and generate results. Click the *train an object detection model* button and choose a training dataset to use. You will be taken to a page that allows you to modify the settings used to train the model.

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	ver infrared - split	Ready for use	mAP=55.685%		
spire by image	rimage	B Report	O View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	0 analyses
	ver infrared - split video	Ready for use	mAP=51.428% © View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	4 analyses
	ver infrared - split v camera	Ready for use	mAP=45.556% • View predictions	Aug. 17, 2021, 3:54 a.m. <b>by</b> gfrench	0 analyses

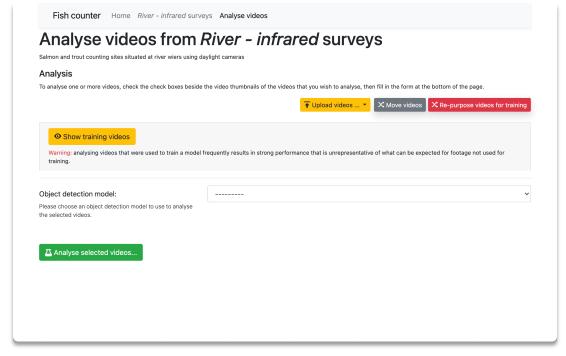
# Using AVIMS

6. Analyse videos

The blue Analyse videos button In Section 6 will take you to the analysis page

Name	Dataset	Status	Accuracy	Created	Analysis results
River infrared	River infrared - split by image	Ready for use	mAP=55.685% • View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	0 analyses
River infrared - split by video	River infrared - split by video	Ready for use	mAP=51.428% • View predictions	Aug. 17, 2021, 3:55 a.m. <b>by</b> gfrench	4 analyses
River infrared	River infrared - split by camera	Ready for use	mAP=45.556% © View predictions	Aug. 17, 2021, 3:54 a.m. <b>by</b> gfrench	0 analyses
Train an object detec	ction model 👻				
Train an object detec	stion model 🔻				
Analyse videos		section 5 above to an	alyse videos and generate	reports.	
Analyse videos e one of the object det			, ,	reports.	
Analyse videos e one of the object det	ection models listed in s		, ,	reports.	
Analyse videos e one of the object det cking the <i>analyse video</i> • Upload videos for ar	ection models listed in s os button will take you to nalysis	o a page where you c	, ,		

If you haven't uploaded any videos for analysis yet, it will look like this:



#### If you have, it will look like this:

Fish counter Home Watery things surveys Analyse videos Admin

#### Analyse videos from *Watery things* surveys

Things that live in the water

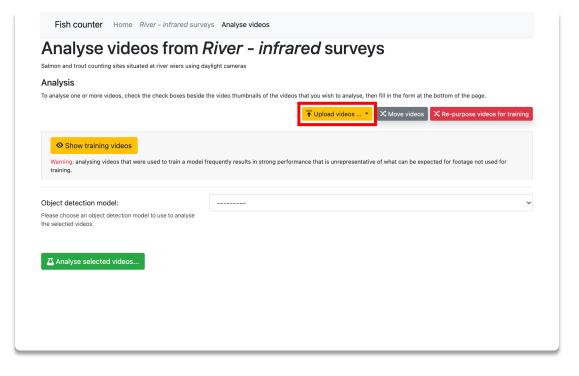
#### Analysis

To analyse one or more videos, check the check boxes beside the video thumbnails of the videos that you wish to analyse, then fill in the form at the bottom of the page.

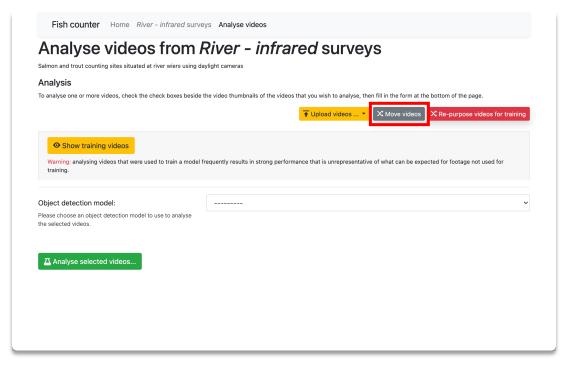
<b>↑</b> Upload vid	eos 🔻 🗙 Move videos 🗙 Re-purpose videos for training
Original filename: 30. DL 131219_144903 fish Down.mp4, Size: 1020.3 KB Uploaded: May 16, 2021, 6:49 a.m. by admin Resolution: 1280 x 720, FPS: 10.0, Frame count: 50, Length: 0:00:05 Size: 884.4 K8 (after re-encoding) Camera: Central	Most recent analysis: July 28, 2021, 4:39 p.m.         ▶ Play video         ▲ D/L video         ■ Summary         See:
Original filename: 22. DL - 230419_153023_otter Down.mp4, Size: 1.1 MB Uploaded: July 20, 2021, 9:40 p.m. by admin Resolution: 1280 x 720, FPS: 25.0, Frame count: 150, Length: 0:00:06 Size: 1.1 MB (after re-encoding) Camera: Central	Most recent analysis: July 28, 2021, 4:39 p.m.         ▶ Play video       ± D/L video         E Summary       ±         See:       Image: All results
Original filename: 36. DL 070519_105820_otter-fish down.mp4, Size: 5.3 MB Uploaded: July 20, 2021, 9:41 p.m. by admin Resolution: 1280 x 720, FPS: 10.0, Frame count: 110, Length: 0:00:11 Size: 5.0 MB (after re-encoding) Camera: Central	Most recent analysis: July 28, 2021, 4:39 p.m.         ▶ Play video       ▲ D/L video         E Summary       ▲         See:       Image: All results
Original filename: 38. DL - 070519_195205_otter down.mp4, Size: 2.5 MB Uploaded: July 20, 2021, 9:41 p.m. by admin Resolution: 1280 x 720, FPS: 10.0, Frame count: 130, Length: 0:00:13 Size: 1.7 MB (after re-encoding) Camera: Central	Most recent analysis: July 28, 2021, 4:39 p.m.         ▶ Play video         ▲ D/L video         ■ Summary         See:         Image: All results
 Original filoname: 40 DL _ ch04_20100521055200. fick up mod_ Size: 5.5 MP	Most recent analysis: July 20-2021 4:20 n m

To analyse videos, you will need to upload them

- Click the yellow Upload videos button
- A menu will appear asking you to choose a camera to associate them with
  - If you need to create a new camera, you will need to return to the survey page, then to the upload page as in section 1
- You will be taken to an upload page where you can upload video files

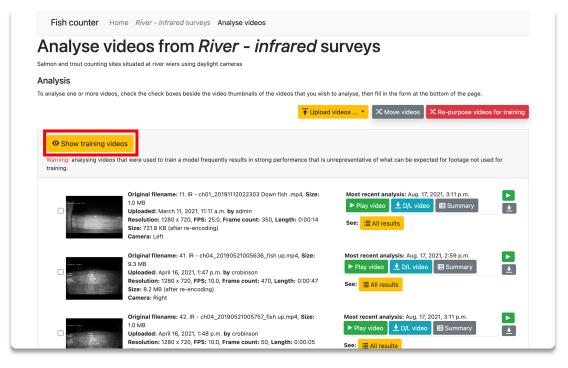


The *Move videos* button allows you to change the camera that videos are associated with, as with the upload page in section 1



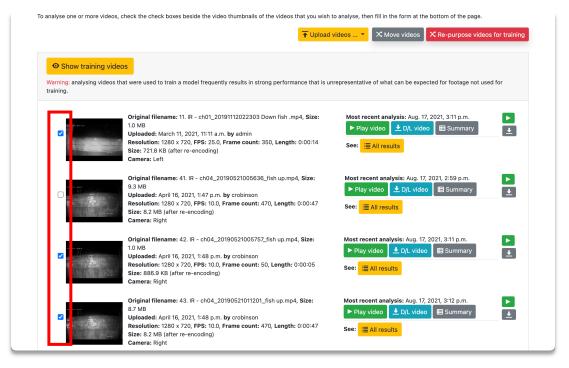
Clicking the yellow *Show training videos* will show the videos that were used for training.

• You can analyse these too, with the caveat that analysing a video that was used for training will get strong performance unrepresentative of what can be expected on other footage



You need to select the videos you want to analyse.

• For each video you want to analyse, tick the checkbox to the left of its' thumbnail image.



Choose the object detection model using the drop-down near the bottom of the page. One is selected here.

Then click the green *Analyse* selected videos... button



The settings in the *analyse videos* form can be left as is.

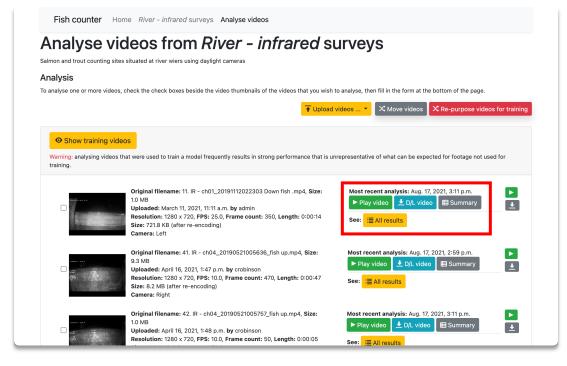
Click the green *Analyse* button to queue the analysis job.

ODO: Give warnings for videos previously inferenced using thi	s model
Videos to analyse	
45. IR - ch04_20190521012926_fish up.mp4	
46. IR - ch04_20190521013420_fish up.mp4	
23. IR - 240419_044704_otter Down.mp4	
25. IR - 270219_235046_otter Down-camera2.mp4	
core threshold	0.5
e detection threshold, between 0 and 1. Objects whose prediction score is less than this lue will; be ignored as false detections.	
acker: minimum detected franes	4
e object must be detected on at least this number of frames to be regarded as present in e video. Objects detected for less than this number of frames are discarded.	
acker: max missing run	3
e object may go missing for up to this number of frames and still be re-detected. An ject that goes missing for longer than this will be lost.	
acker: match threshold	0.25
e overlap threshold between an object tracked in prior frames and a detection in the	

using Pivor infrared - split by vide

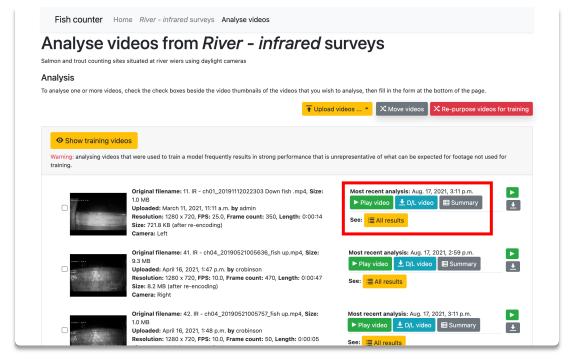
Fish counter Home River - infrared surveys Analyse videos

Once at least one analysis of a video is complete (a video can be analysed multiple times, e.g. by multiple models), the results can be quickly accessed from the analysis page.

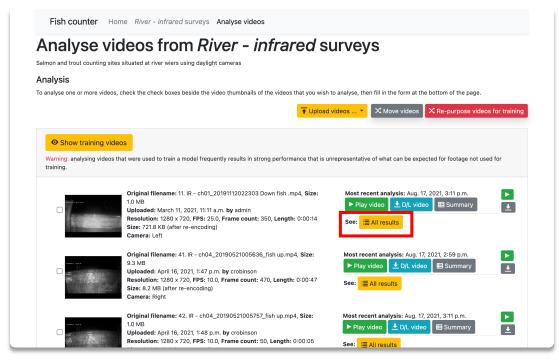


A video with detected and tracked objects can be played or downloaded.

A tabular summary of the detections can be downloaded as a .CSV file or viewed as a table.

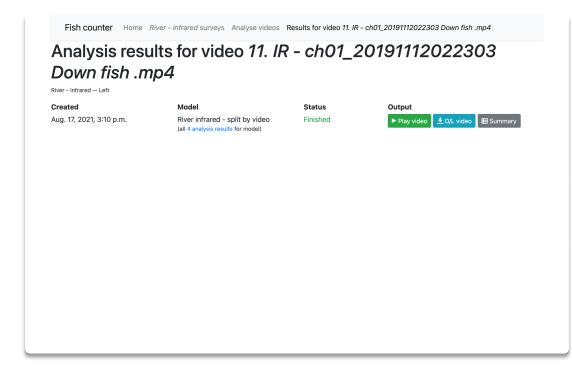


Clicking the yellow *All results* button will show you all the analysis results for that video...



In this case, this video has only been analysed once, so only one entry is visible.

Clicking the small blue '4 analysis results' in the Model column will show all the analysis results created using that model...

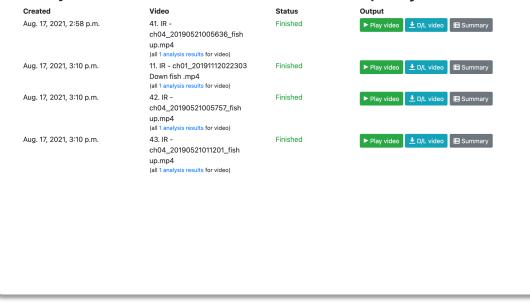


Analysis results from a model

This can also be accessed from the survey page; the last column in the table in section 5 links here.

Fish counter Home River - infrared surveys Analysis results from model River infrared - split by video

#### Analysis results from model River infrared - split by video



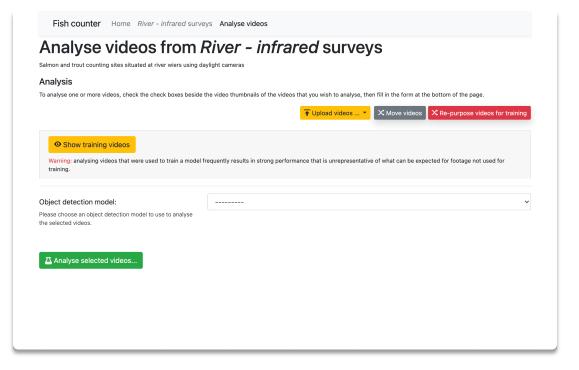
# Using AVIMS

Circular work-flow

#### Circular work-flow

Let's assume that after watching the video generated by analysis, you conclude that the performance needs improvement.

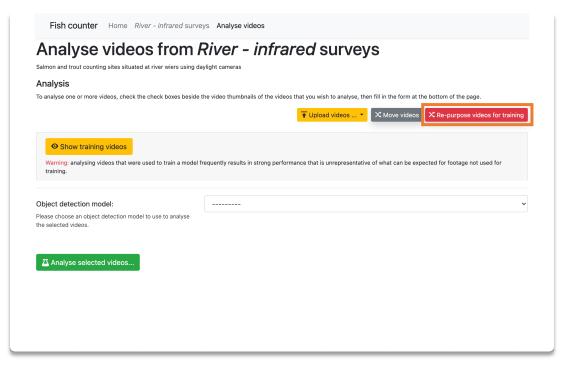
One option is to add this video to the training set.



### Circular work-flow

Adding an analysed video to the training set

- Click the red *Re-purpose videos* for training button
- This will take you to a page where you can select videos to move to the training section



#### Circular work-flow

Adding an analysed video to the training set

- Return to the upload footage page (section 1) and choose frames for labelling
- Label these frames (section 3)
- Repeat the subsequent steps...

# Thank you!

I hope you found it helpful

#### Contact

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