

AND OTHER ANIMALS

Counting Koalas with Kafka

 @SimonAubury

 @saubury

 /thoughtworks





Simon Aubury

Principal Data Engineer



Kafka enthusiast



Confluent Community Catalyst



Sydney, Australia



Baz

Southern koala



Endangered



Eucalyptus trees



Australia

 **thoughtworks**

@SimonAubury

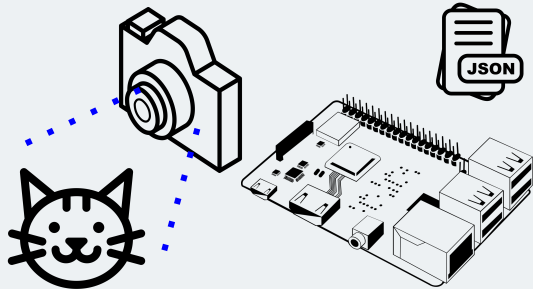
A photograph of a dense forest with tall, slender trees. The trees have light-colored, peeling bark, characteristic of eucalyptus. The foliage is lush green, and the sunlight filters through the canopy, creating a dappled light effect on the forest floor.

How many koalas are here?

Is this less than yesterday?

What's changed?

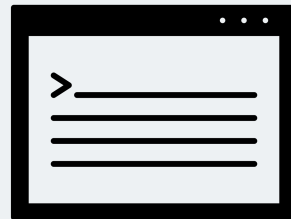
Photo by [David Clode](#) on [Unsplash](#)



TensorFlow /
Raspberry Pi camera



Console



1. Kafka-Cat



Raspberry Pi

Cameras in the wilderness

Raspberry Pi is a **low cost** credit-card sized computer

With an attached camera they have sufficient processing power for edge ML detection with TensorFlow Lite



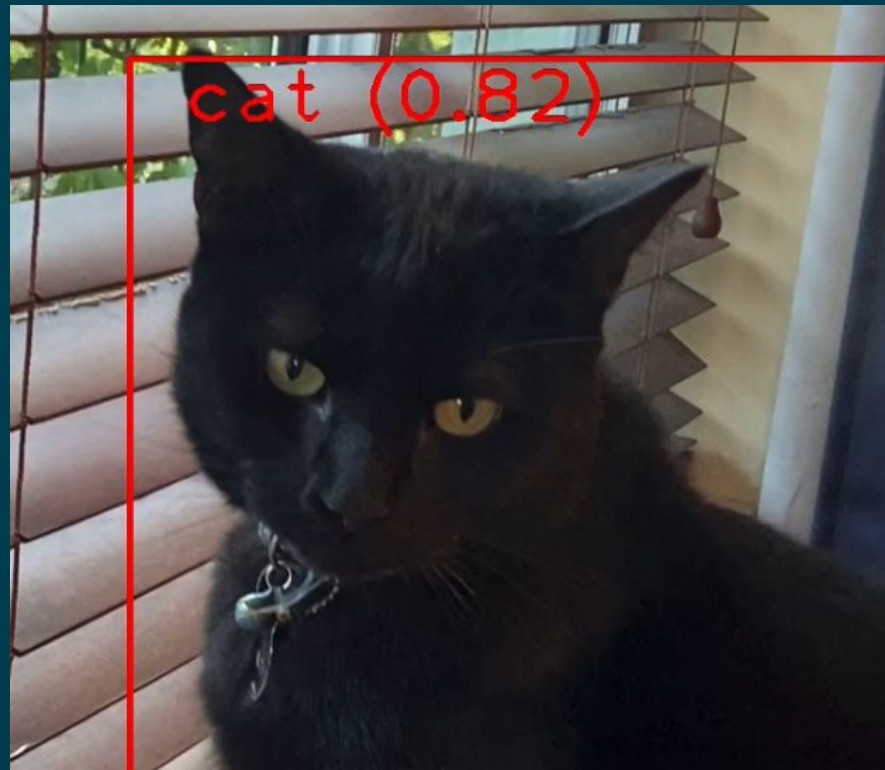
TensorFlow

TensorFlow Lite object detection

Identify which of a known set of objects might be present and provide information about their positions within the image.

```
{ "cat": 0.82 }
```

```
{"cat":1, "koala":0}
```



Open Camera

```
# Start capturing video input from the camera
```

```
cap = cv2.VideoCapture()
```

Load model

```
# Load model efficientdet_lite0.tflite, set threshold
```

```
detector = vision.ObjectDetector.create_from_options(70%)
```

Single image

```
# Continuously capture camera images
```

```
while True:
```

```
    image = cap.read()
```

Object detect

```
# Run object detection estimation using the model
```

```
detection_result = detector.detect(image)
```

Kafka produce

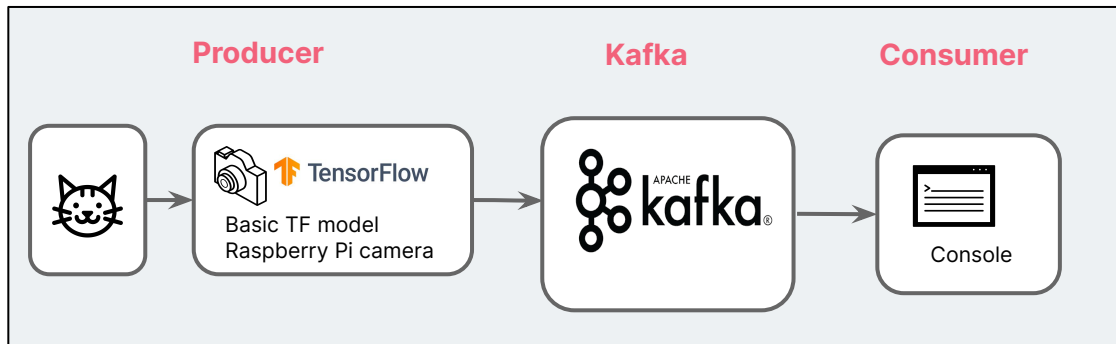
```
# Produce to objects topic
```

```
kafka_producer.produce( detection_result )
```

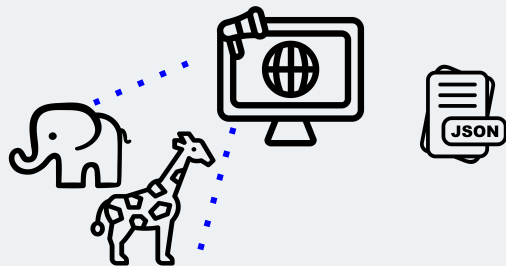
<https://github.com/saubury/wildlife-watch/blob/main/detect.py>

Architecture

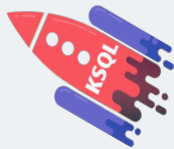
MVP version



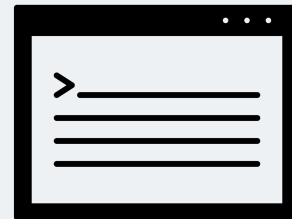
```
-zsh
Python  #1 -zsh
saubury:wildlife-watch % kafka-console-consumer --bootstrap-server localhost:9092 --topic objects
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.74}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.74}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.74}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.76}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.76}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.76}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.72}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.76}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.72}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.72}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.71}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.71}], "objects_count": {"cat": 1}}
{"camera_name": "raspberrypi", "objects_found": [{"class_name": "cat", "probability": 0.71}], "objects_count": {"cat": 1}}
```

TensorFlow / Zoo
Livestream



Console



2. Zoo-keeping with ksqlDB

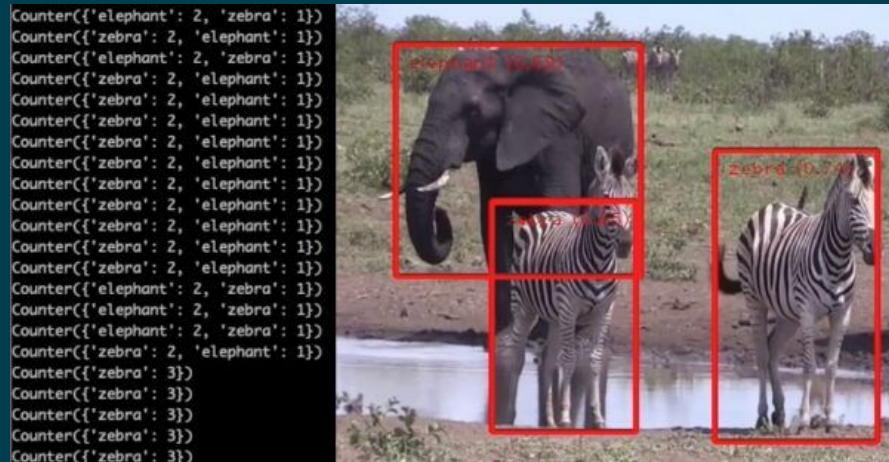


Zoo camera feed

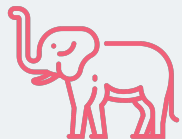
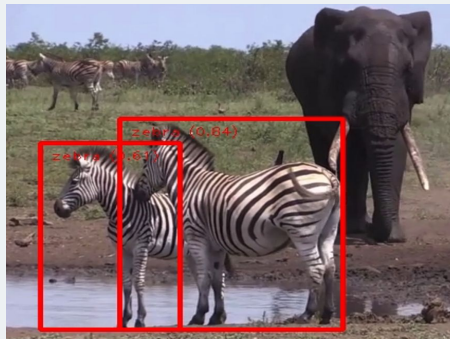
Expanding the animal collection

More animals from the zoo

- Additional Kafka producer
- Laptop based
- Source video webcam feed from local zoo (Pyautogui & OpenCV)



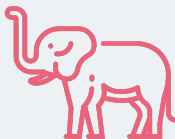
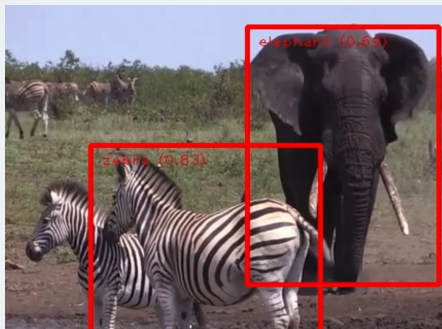
```
{
  "camera_name": "zoo-webcam",
  "objects_count": {
    "elephant": 1,
    "zebra": 2
  }
}
```



0



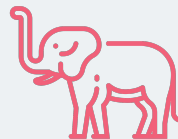
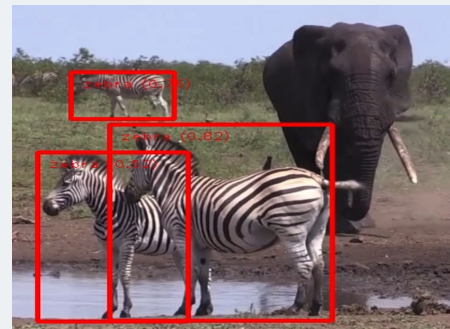
2



1



1



0



3

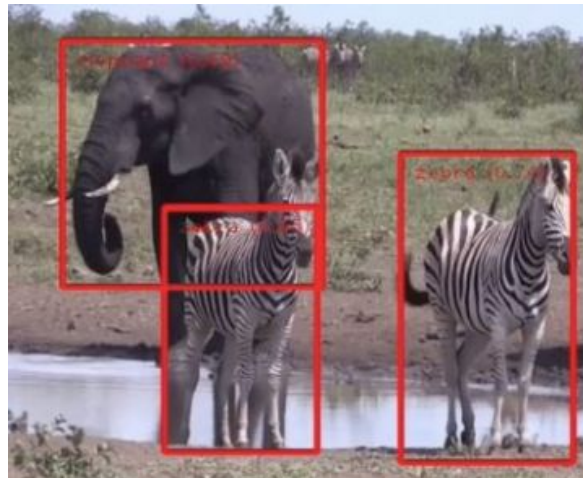
Payload extraction

Payload

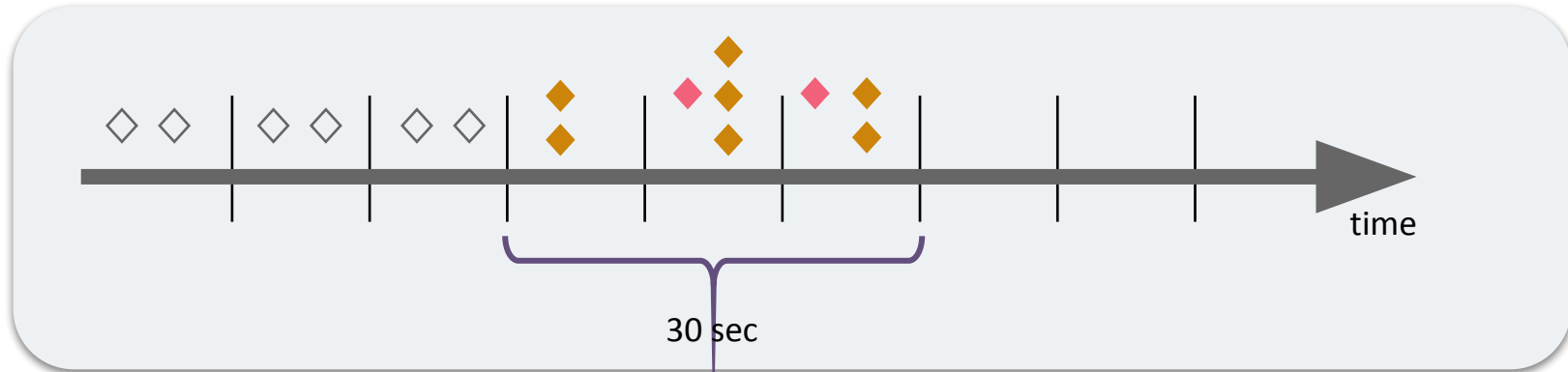
```
{  
  "camera_name": "zoo-webcam",  
  "objects_count": {  
    "elephant": 1,  
    "zebra": 2  
  }  
}
```

KSQL pivot

```
create stream animals as  
select  extractjsonfield(objects_count, '$.elephant') as elephant  
        , extractjsonfield(objects_count, '$.zebra') as zebra  
        , < many more animals >  
from objects;
```



ksqlDB



KSQl table

```
create table zoo as
select max(elephant) , max(zebra)
from animals window tumbling (size 30 seconds)
group by camera_name
emit changes;
```



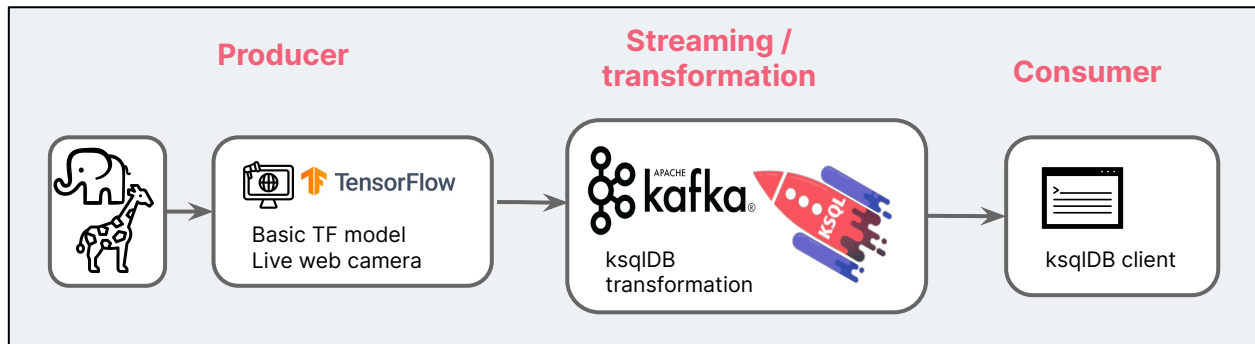
1



3

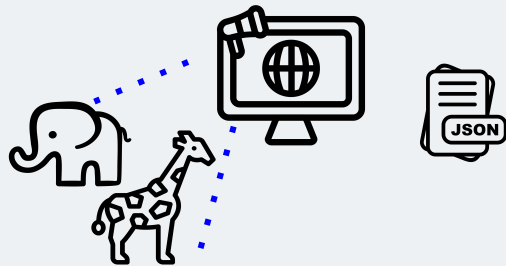
Architecture

Zoo-keeper version

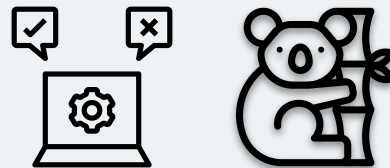


```
ksql> select camera_name, max_elephant, max_zebra from zoo emit changes;
```

CAMERA_NAME	MAX_ELEPHANT	MAX_ZEBRA
lmylaptop	null	1
lmylaptop	null	2
lmylaptop	null	3
lmylaptop	1	3
lmylaptop	1	3
lmylaptop	1	3

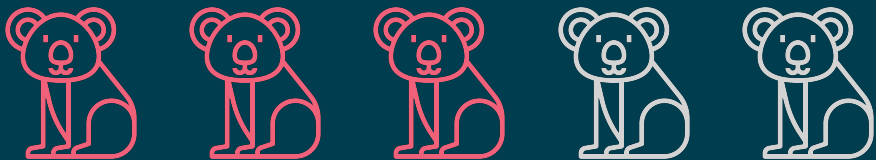


TensorFlow / Zoo
Livestream



Transfer learning /
koala training data

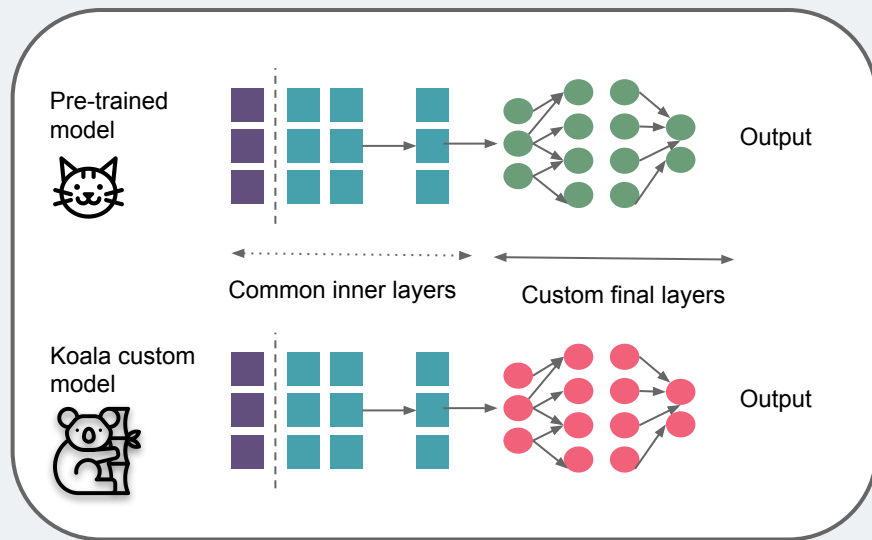
3. Transfer learning with Koalas



Transfer learning

Custom Koala object detection model

i **Transfer learning** is an ML technique that focuses on using knowledge gained while solving one problem ... and applying it to a different but related problem.

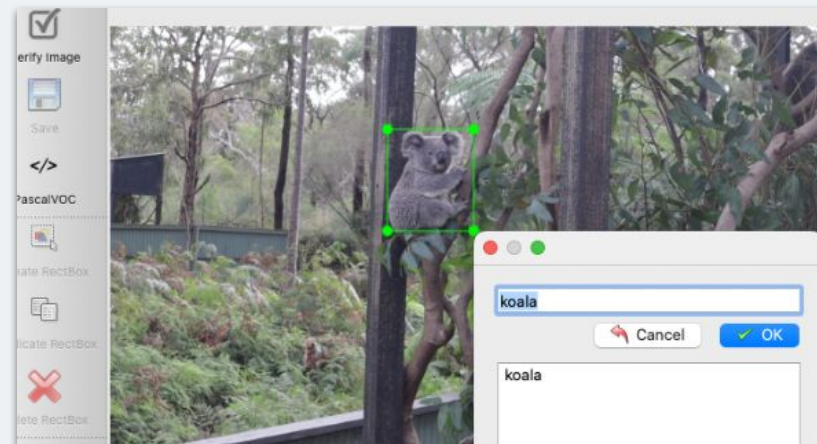


Transfer learning

Custom Koala object detection model

⚠ EfficientDet TensorFlow Lite trained on COCO 2017 dataset ... over 200K labeled images

🎯 Goal - retrain an model with koala dataset to train a custom object detection model



Edges, shapes,
outlines,
contrasts



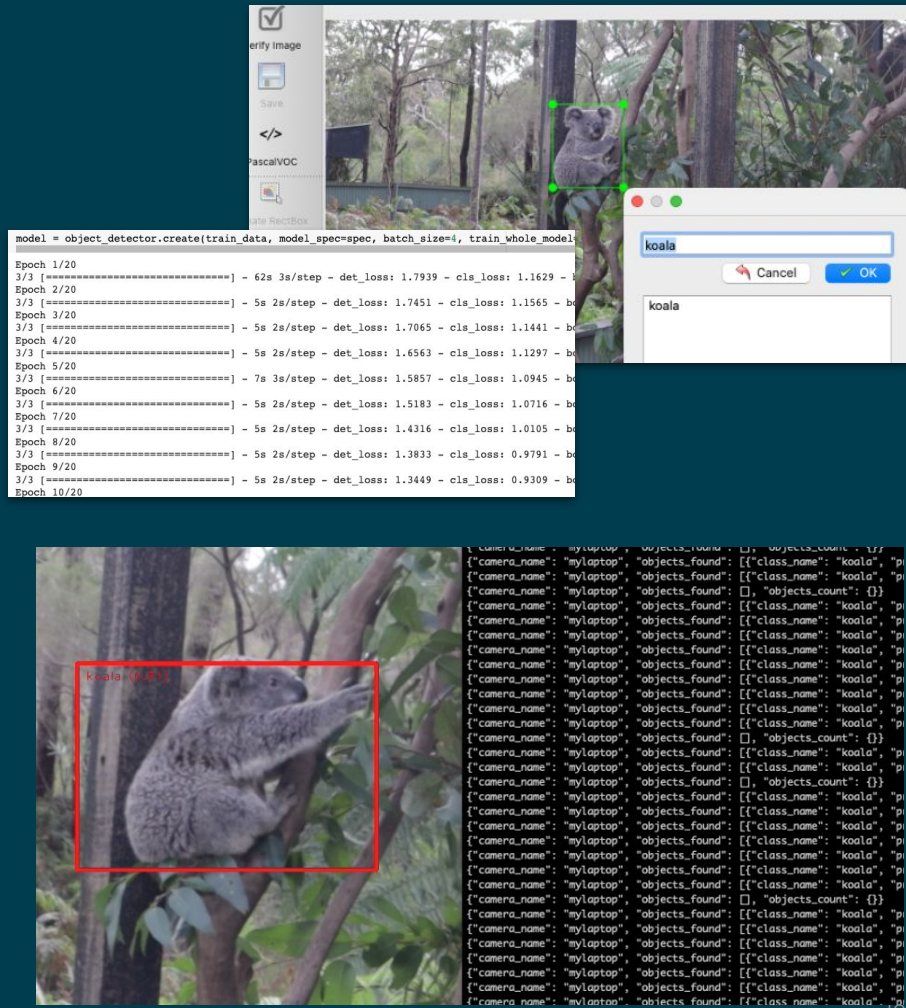
Koala specific
data set

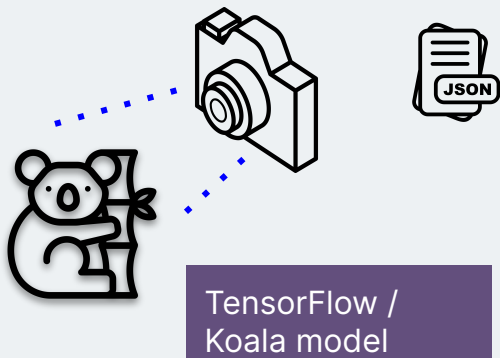
Koala model

Retraining a TensorFlow Lite model

TensorFlow Lite Model Maker

- Koala dataset with **Labellmg**
- Train the TensorFlow model
- Export as a TensorFlow Lite model.
- Evaluate model
- Deploy model to RaspberryPi





Kafka connect /
elastic sink



Kafka connect /
http sink



4. Analysis and alerting

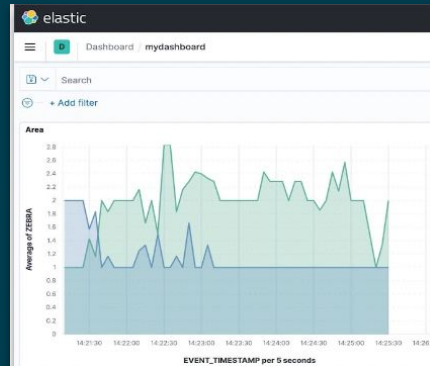


Analysis and alerting

Command line is cool, but ...

- Real-time dashboard with **Kibana**
- Phone alerts with **Telegram**

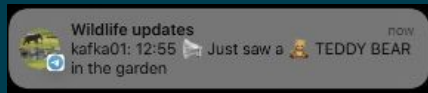
Kafka Connect is a framework for connecting Kafka with external systems



Dashboard



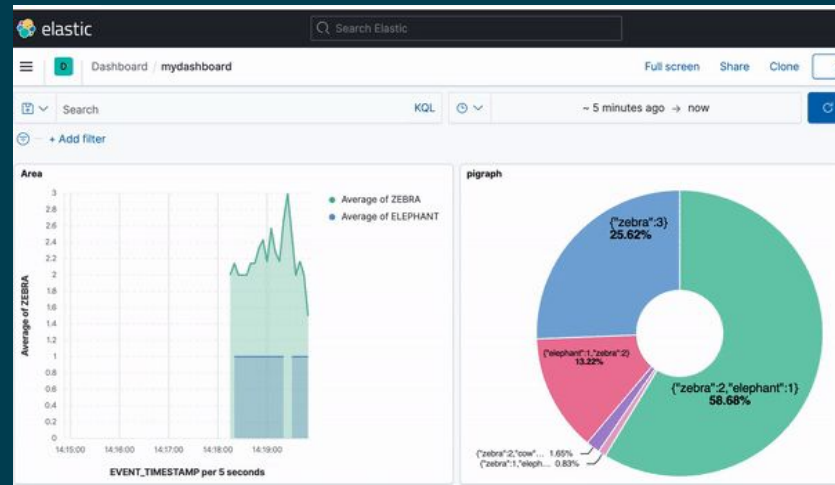
Alerts



Kibana dashboard



Kafka Connect with Kafka Connect Elasticsearch connector to send both the **animals** and **zoo** Kafka topics to Elasticsearch indexes.



Building the dashboard

Kafka connect / elastic sink

Kafka Topic

Elastic sink class

Index name

Timestamp

```
{
  "topics": "ZOO",
  "key.converter": "org.apache.kafka.connect.storage.StringConverter",
  "value.converter.schemas.enable": "false",
  "connector.class": "io.confluent.connect.elasticsearch.ElasticsearchSinkConnector",
  "key.ignore": "true",
  "value.converter": "org.apache.kafka.connect.json.JsonConverter",
  "type.name": "type.name=kafkaconnect",
  "topic.index.map": "ZOO:zoo",
  "connection.url": "http://elasticsearch:9200",
  "transforms": "ExtractTimestamp",
  "transforms.ExtractTimestamp.type": "org.apache.kafka.connect.transforms.InsertField$Value",
  "transforms.ExtractTimestamp.timestamp.field": "EVENT_TIMESTAMP"
}
```

ZOO*

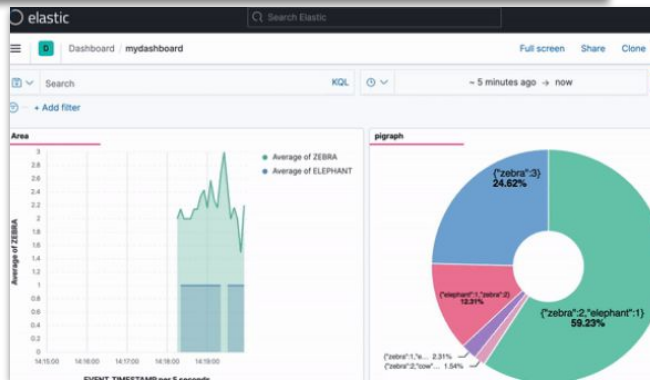
Time field: 'EVENT_TIMESTAMP'

This page lists every field in the **zoo*** index and the field's associated core type as recorded by Elasticsearch Mapping API

Fields (8) Scripted fields (0) Field filters (0)

Search

Name	Type	Format	Searchable
EVENT_TIMESTAMP	date		•
MAX_ELEPHANT	number		•
MAX_ZEBRA	number		•

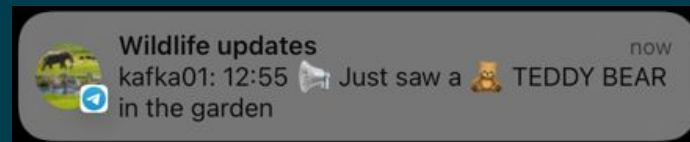


Telegram bot



Created wildlife Telegram bot with exposed **HTTP-based** interface

Telegram bot alert for each record in **teddybear-telegram-topic** Kafka



Teddy bear alerts

Kafka connect / http sink

KSQL stream

```
create stream teaddybear-telegram-topic
as
select '🐻 Just saw a 🐻 TEDDY BEAR in the garden' as message
from animals
where teddybear > 0 ;
```

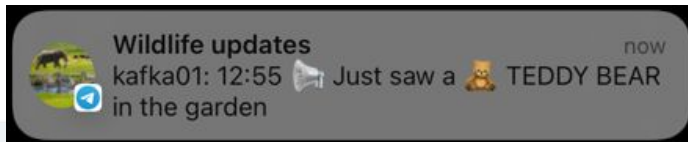
Kafka topic

Telegram API

Chat ID

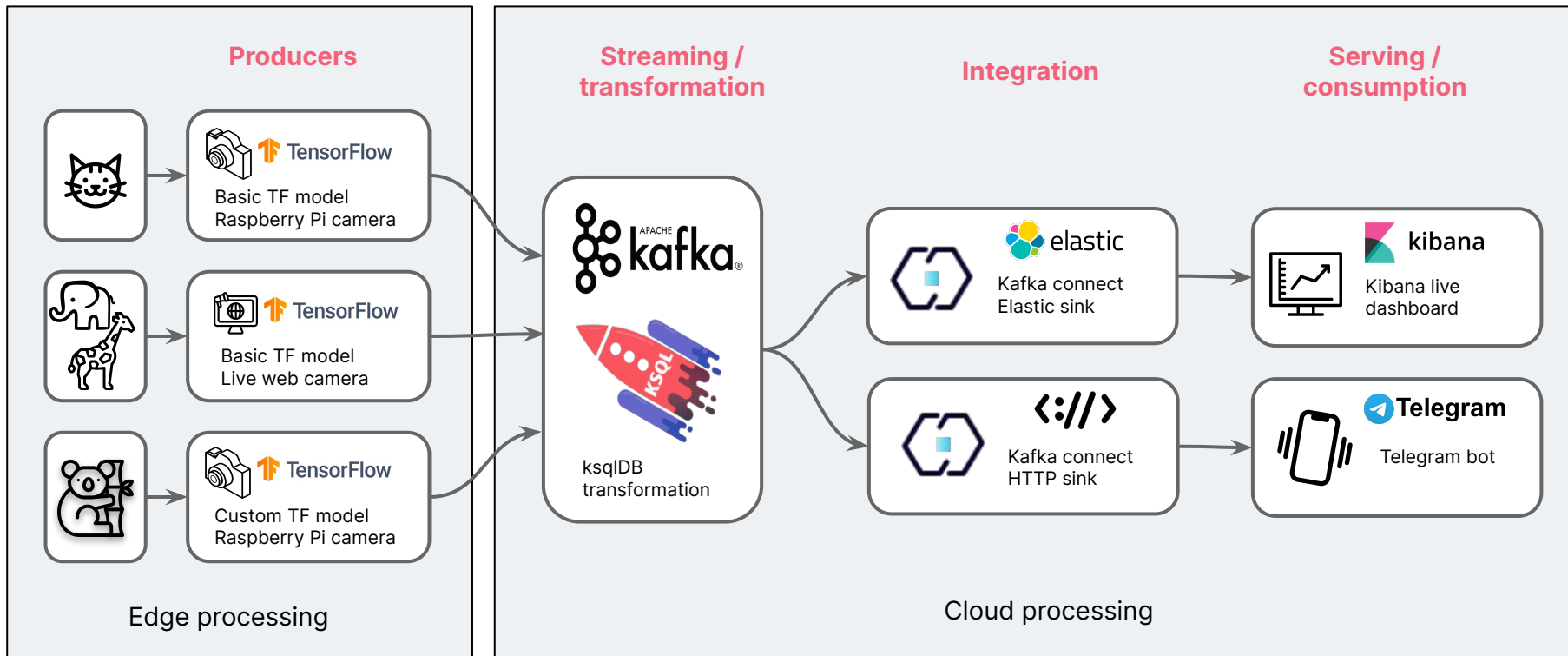
Regular expression

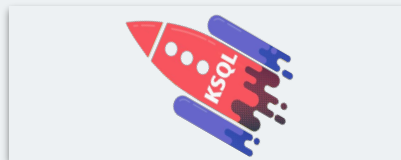
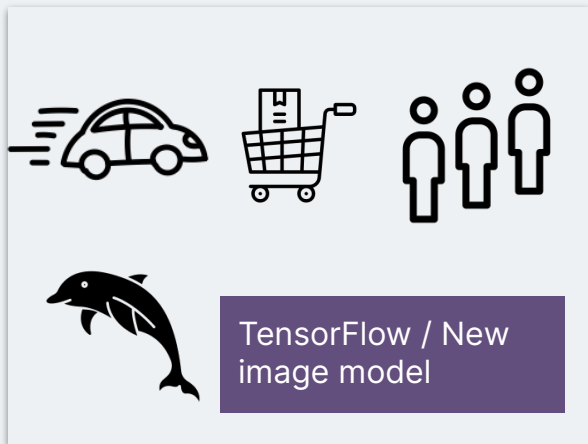
```
{
  "topics": "teaddybear-telegram-topic",
  "input.data.format": "JSON",
  "connector.class": "HttpSink",
  "confluent.topic.bootstrap.servers": "broker:29092",
  "confluent.topic.replication.factor": "1",
  "reporter.bootstrap.servers": "broker:29092",
  "reporter.result.topic.name": "success-responses",
  "reporter.result.topic.replication.factor": "1",
  "reporter.error.topic.name": "error-responses",
  "reporter.error.topic.replication.factor": "1",
  "http.api.url": "https://api.telegram.org/botXXXXXX/sendMessage",
  "request.method": "POST",
  "headers": "Content-Type: application/json",
  "request.body.format": "string",
  "batch.max.size": "1",
  "batch.prefix": "{\"chat_id\":\"-123456\",",
  "batch.suffix": "}",
  "regex.patterns": ".*MESSAGE=(.*)",
  "regex.replacements": "\"text\":\"$1\"",
  "regex.separator": "~",
  "tasks.max": "1",
  "value.converter": "org.apache.kafka.connect.storage.StringConverter",
  "name": "teaddybear-telegram-sink"
}
```



Architecture

Final version

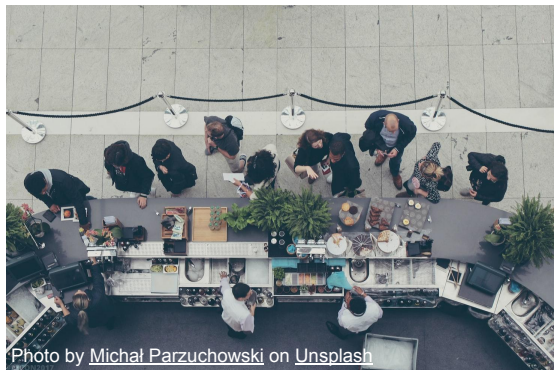




5. How does this help me?



Events are everywhere



Coffee queue **wait times**



Shopping trolley **usage**



Car park **occupancy**

Understand **streams that matter** - object detection plus stream processing

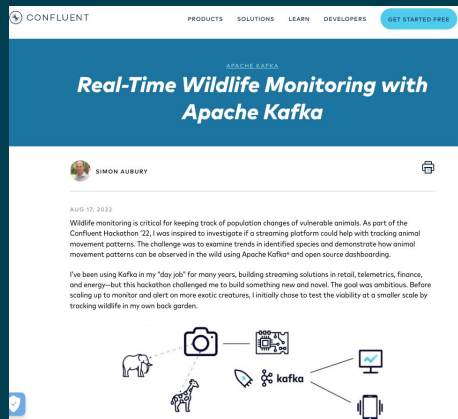


That which is measured improves.

*That which is measured and
reported improves exponentially*

Karl Pearson

Thanks / Any questions?



/thoughtworks

@SimonAubury

Photo by [David Clode](#) on [Unsplash](#)

