# Vision processing / with Kafka & ksqlDB

Can I arrange my sock drawer using machine learning?

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I am here because I love streaming



# Pairing socks with ML

Can I arrange my sock drawer with transfer learning to build a custom sock image classification model?



github.com/saubury/socksort







# 

### Events are everywhere

Event-driven architecture is an architecture paradigm promoting the production, detection, consumption of and reaction to events.

Object detection is a stream of events

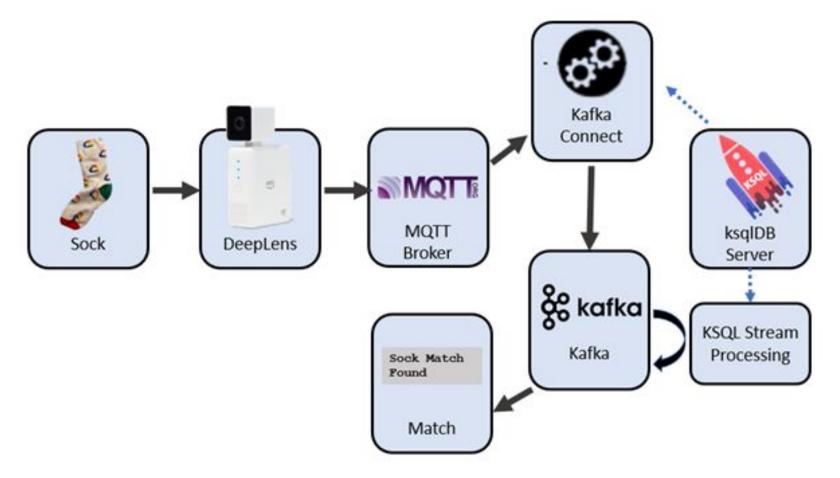






**Events are very messy** What I think will happen What actually time No sock No sock happens time No sock Phantom sock







#### Let's start at the finish

- Hold a sock in front of camera
- Classification for each frame
- Messages written to MQTT
- Messages transported to Kafka
- Stream processing on Kafka
- Socks are matched





# Deep learning video camera AWS Deeplens

#### **AWS Deeplens Hardware**

- A "deep-learning enabled video camera".
  - 4MP video camera
  - Intel Atom Processor
  - o 8GB RAM
  - o Runs Ubuntu
- Plenty of hardware to help sort my socks.





# Many photos of socks

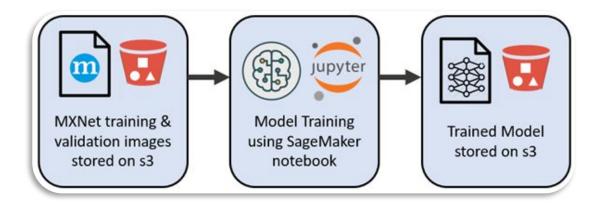
#### Supervised learning image classification

- Requires training data
- Prepare a set of training images
- I need to take a lot of photos of socks







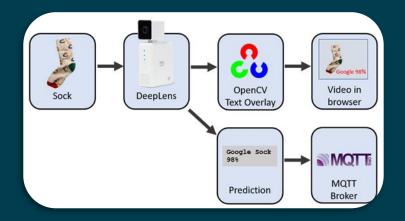


## **Model Training**

Image classification of socks using transfer learning mode.

- Use AWS Sagemaker image classification algorithm in transfer learning
- Deploy a temporary classifier to test the inference function
- Test a few demonstration images can be correctly classified





# Inference Lambda Function

#### On the DeepLens

- Run all images captured by the camera through the classification model.
  - Review a live camera feed within as web-browser
  - OpenCV adds text overlaid on the image.
- Write to a MQTT topic

#### **Inference Lambda Function**

```
while doInfer:
   # Get a frame from the video stream
   ret, frame = awscam.getLastFrame()
   # Raise an exception if failing to get a frame
   if ret == False:
       raise Exception("Failed to get frame from the stream")
   # Resize frame to fit model input requirement
   frameResize = cv2.resize(frame, (input_width, input_height))
   # Run model inference on the resized frame
   inferOutput = model.doInference(frameResize)
   # Output inference result to the fifo file so it can be viewed with mplayer
   parsed_results = model.parseResult(model_type, inferOutput)
   top_k = parsed_results[model_type][0:topk]
   sock_label = labels[top_k[0]["label"]]
   sock_prob = top_k[0]["prob"]*100
   # Write to MOTT
   ison payload = {"image" : sock label, "probability" : sock prob}
   client.publish(topic=iot_topic, payload=json.dumps(json_payload))
   # Write to image buffer; screen display
   msg_screen = '{} {:.0f}%'.format(sock_label, sock_prob)
   cv2.putText(frame, msg_screen, (20,200), cv2.FONT_HERSHEY_SIMPLEX, 5, (0, 0, 255), 12)
   local_display.set_frame_data(frame)
```



## What is MQTT?

- MQTT is lightweight TCP/IP protocol
  - Small footprint
  - Low power
- MQTT acts more like a key/value store

```
mosquitto_sub -h ${MQTT_HOST} -p ${MQTT_PORT} -u ${MQTT_USER} -P
${MQTT_PASS} -t sockfound

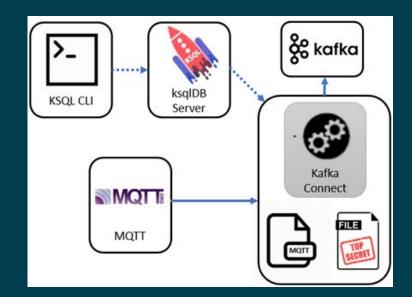
{"image": "Blank", "probability": 37.59765625}
{"image": "Blank", "probability": 41.162109375}
{"image": "Google", "probability": 97.314453125}
{"image": "Google", "probability": 94.970703125}
{"image": "Google", "probability": 64.6484375}
{"image": "Blank", "probability": 67.3828125}
{"image": "Blank", "probability": 50.634765625}
```



# Kafka, Kafka Connect & ksqlDB

Kafka is a distributed streaming platform

- Kafka platform for handling real-time data
- Kafka Connect framework for streaming data between Kafka and other data systems
- ksqIDB build real-time systems with SQL statements



```
CREATE SOURCE CONNECTOR `mqtt-source` WITH(
"connector.class"='io.confluent.connect.mqtt.MqttSourceConnector',
"mqtt.server.uri"='tcp://something.example.com:14437',
"mqtt.username"='some-user',
"mqtt.password"='my-password',
"mqtt.topics"='sockfound',
"kafka.topic"='data_mqtt',
);
```



# **Ghost socks**

Image classifier identifies 3-4 images each second.









# **Event Stream processing**

Goal: find similar messages within a windows of time



```
create table sock_stream_smoothed as
select image
, timestamptostring(windowstart(), 'hh:mm:ss') as last_seen
, windowstart() as window_start
from sock_stream
window tumbling (size 5 seconds)
group by image having count(*) > 3
emit changes;
```

# Window Hopping with ksqlDB

Goal: find 4 or more identical socks in a rolling 5 second window



#### Goal: find pairs of identical socks

```
select image
, case when (count(*)/2)*2 = count(*) then 'Pair' else 'Un-matched'
end as pair_seen
, count(*) as number_socks_seen
from sock_stream_smoothed
group by image
emit changes;
```



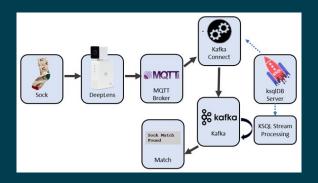
#### Demo

```
-- Creata stream for the MQTT topic
create stream sock_stream(image varchar, probability double)
with (kafka_topic='data_mqtt', value_format='json');
-- Bucket sock images into windows of 5 seconds
create table sock_stream_smoothed as
select image
, timestamptostring(windowstart(), 'hh:mm:ss') as last_seen
, windowstart() as window_start
from sock_stream
window tumbling (size 5 seconds)
where image != 'blank'
group by image having count(*) > 3
emit changes;
-- Find pairs of socks (socks appearing in even numbers)
, case when (count(*)/2)*2 = count(*) then 'Pair' else 'Un-matched' end as pair_seen
, count(*) as number_socks_seen
from sock_stream_smoothed
group by image
emit changes;
```



#### What did I learn?

- AWS Deeplens is super cool
- Object detection is a stream of events
- Events are very messy
- IoT architecture with Kafka is really scalable
- Stream processing with ksqlDB is very convenient
- Sock sorting does not make you interesting at parties







# Q&A





https://github.com/saubury/socksort



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