

```
[ ]: import tensorflow as tf
      from pathlib import Path
      tf_download_root = "http://download.tensorflow.org/data/"
      filename = "quickdraw_tutorial_dataset_v1.tar.gz"
      filepath = tf.keras.utils.get_file(filename,
                                          tf_download_root + filename,
                                          cache_dir=".",
                                          extract=True)

      quickdraw_dir = Path(filepath).parent

[ ]: train_files = sorted(
      [str(path) for path in quickdraw_dir.glob("training.tfrecord-*")]
    )[:3]
      eval_files = sorted(
      [str(path) for path in quickdraw_dir.glob("eval.tfrecord-*")]
    )

[ ]: import numpy as np
      import matplotlib.pyplot as plt

[ ]: with open(quickdraw_dir / "eval.tfrecord.classes") as test_classes_file:
      test_classes = test_classes_file.readlines()

      with open(quickdraw_dir / "training.tfrecord.classes") as train_classes_file:
      train_classes = train_classes_file.readlines()

      assert train_classes == test_classes
      class_names = [name.strip().lower() for name in train_classes]
      sorted(class_names)

      def parse(data_batch):
          feature_descriptions = {
              "ink": tf.io.VarLenFeature(dtype=tf.float32),
              "shape": tf.io.FixedLenFeature([2], dtype=tf.int64),
```

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    "shape": tf.io.FixedLenFeature([2], dtype=tf.int64),
    "class_index": tf.io.FixedLenFeature([1], dtype=tf.int64)
}
examples = tf.io.parse_example(data_batch, feature_descriptions)
flat_sketches = tf.sparse.to_dense(examples["ink"])
sketches = tf.reshape(flat_sketches, shape=[tf.size(data_batch), -1, 3])
lengths = examples["shape"][:, 0]
labels = examples["class_index"][:, 0]
return sketches, lengths, labels

def quickdraw_dataset(filepaths, batch_size=32, shuffle_buffer_size=None,
                      n_parse_threads=5, n_read_threads=5, cache=False):
    dataset = tf.data.TFRecordDataset(filepaths,
                                     num_parallel_reads=n_read_threads)

    if cache:
        dataset = dataset.cache()
    if shuffle_buffer_size:
        dataset = dataset.shuffle(shuffle_buffer_size)
    dataset = dataset.batch(batch_size)
    dataset = dataset.map(parse, num_parallel_calls=n_parse_threads)
    return dataset.prefetch(1)

train_set = quickdraw_dataset(train_files, shuffle_buffer_size=10000)
valid_set = quickdraw_dataset(eval_files[:5])
test_set = quickdraw_dataset(eval_files[5:])

def draw_sketch(sketch, label=None):
    origin = np.array([[0., 0., 0.]])
    sketch = np.r_[origin, sketch]
    stroke_end_indices = np.argwhere(sketch[:, -1]==1.)[:, 0]
    coordinates = sketch[:, :2].cumsum(axis=0)
    strokes = np.split(coordinates, stroke_end_indices + 1)
    title = class_names[label.numpy()] if label is not None else "Try to guess"
    plt.title(title)
    plt.plot(coordinates[:, 0], -coordinates[:, 1], "y:")

```

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for stroke in strokes:
    plt.plot(stroke[:, 0], -stroke[:, 1], "-.")
plt.axis("off")

def draw_sketches(sketches, lengths, labels):
    n_sketches = len(sketches)
    n_cols = 4
    n_rows = (n_sketches - 1) // n_cols + 1
    plt.figure(figsize=(n_cols * 3, n_rows * 3.5))
    for index, sketch, length, label in zip(range(n_sketches), sketches, lengths, labels):
        plt.subplot(n_rows, n_cols, index + 1)
        draw_sketch(sketch[:length], label)
    plt.show()

for sketches, lengths, labels in train_set.take(1):
    draw_sketches(sketches, lengths, labels)

def crop_long_sketches(dataset, max_length=100):
    return dataset.map(lambda inks, lengths, labels: (inks[:, :max_length], labels))

cropped_train_set = crop_long_sketches(train_set)
cropped_valid_set = crop_long_sketches(valid_set)
cropped_test_set = crop_long_sketches(test_set)

model = tf.keras.Sequential([
    tf.keras.layers.Conv1D(32, kernel_size=5, strides=2, activation="relu"),
    tf.keras.layers.BatchNormalization(),
    tf.keras.layers.Conv1D(64, kernel_size=5, strides=2, activation="relu"),
    tf.keras.layers.BatchNormalization(),
    tf.keras.layers.Conv1D(128, kernel_size=3, strides=2, activation="relu"),
    tf.keras.layers.BatchNormalization(),
    tf.keras.layers.LSTM(128, return_sequences=True),
    tf.keras.layers.LSTM(128),
    tf.keras.layers.Dense(len(class_names), activation="softmax")
])

```

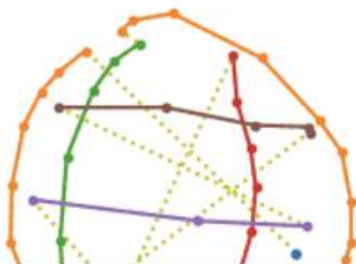
```
optimizer = tf.keras.optimizers.SGD(learning_rate=1e-2, clipnorm=1.)
model.compile(loss="sparse_categorical_crossentropy",
              optimizer=optimizer,
              metrics=["accuracy", "sparse_top_k_categorical_accuracy"])
history = model.fit(cropped_train_set, epochs=1,
                  validation_data=cropped_valid_set)

y_test = np.concatenate([labels for _, _, labels in test_set])
y_probas = model.predict(test_set)

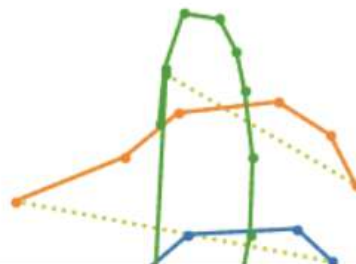
np.mean(tf.keras.metrics.sparse_top_k_categorical_accuracy(y_test, y_probas))

n_new = 10
Y_probas = model.predict(sketches)
top_k = tf.nn.top_k(Y_probas, k=5)
for index in range(n_new):
    plt.figure(figsize=(3, 3.5))
    draw_sketch(sketches[index])
    plt.show()
    print("Top-5 predictions:".format(index + 1))
    for k in range(5):
        class_name = class_names[top_k.indices[index, k]]
        proba = 100 * top_k.values[index, k]
        print("  {}. {} {:.3f}%".format(k + 1, class_name, proba))
    print("Answer: {}".format(class_names[labels[index].numpy()]))
```

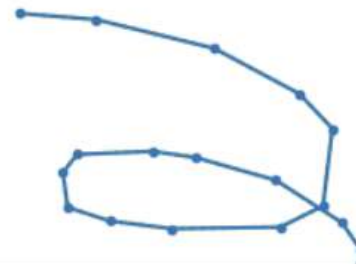
basketball



belt



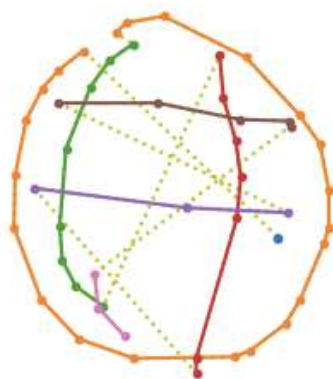
squiggle



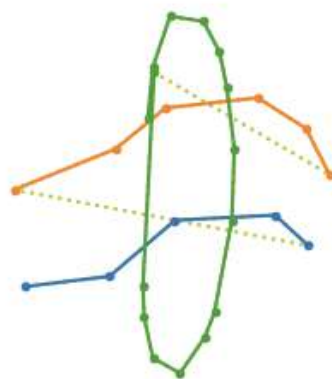
banana



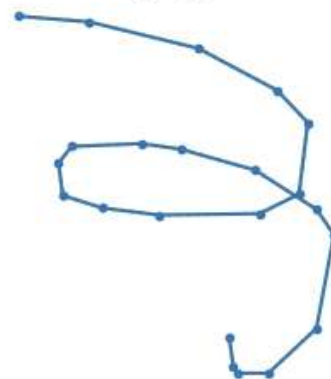
basketball



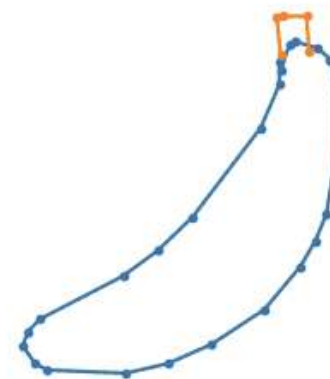
belt



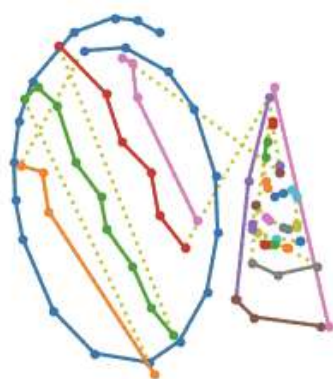
squiggle



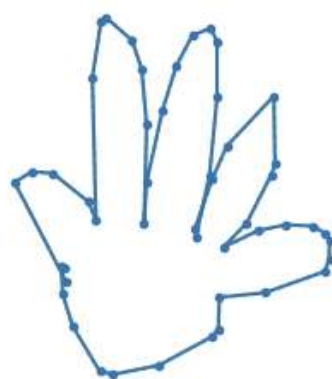
banana



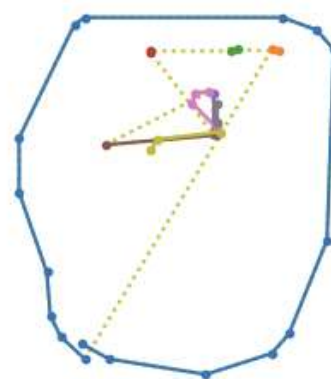
watermelon



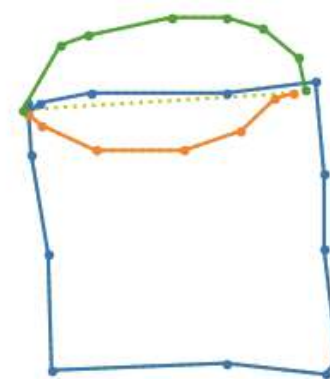
hand



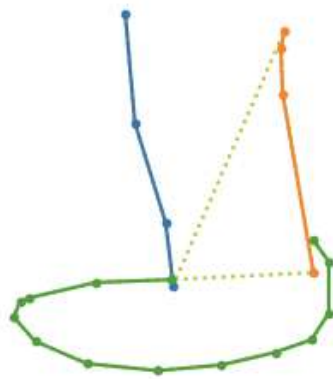
remote control



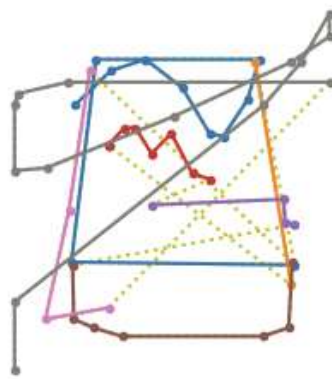
bucket



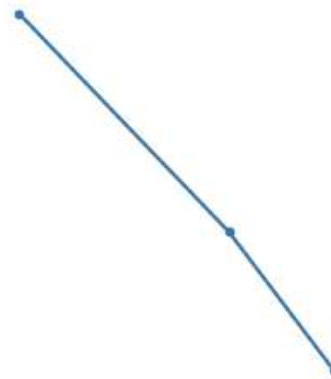
leg



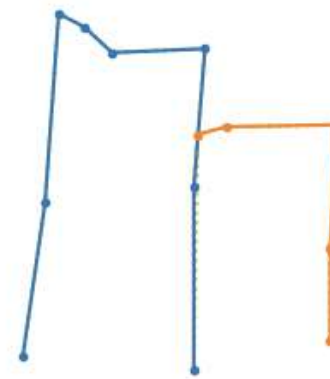
sock



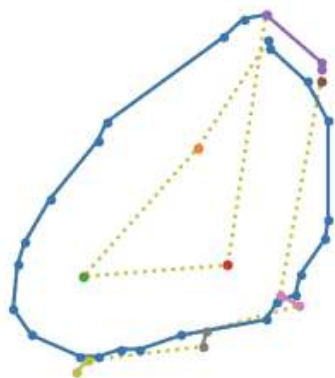
line



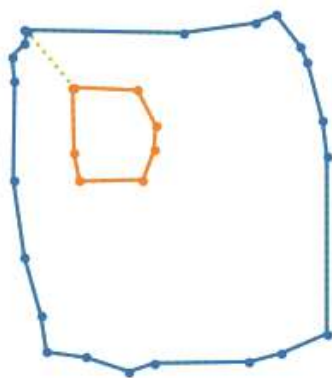
skyscraper



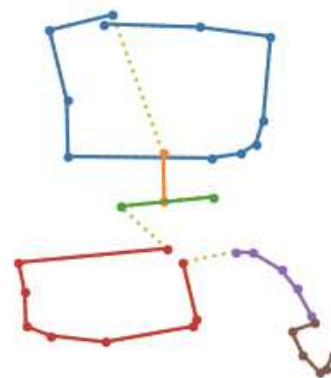
potato



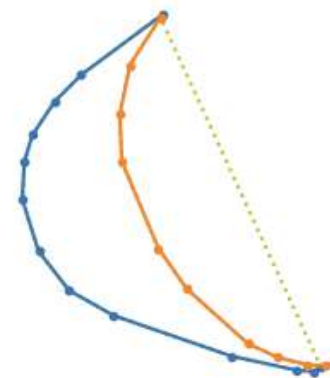
passport



computer



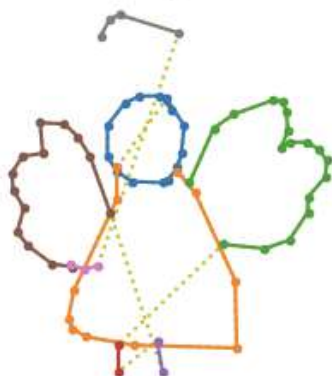
moon



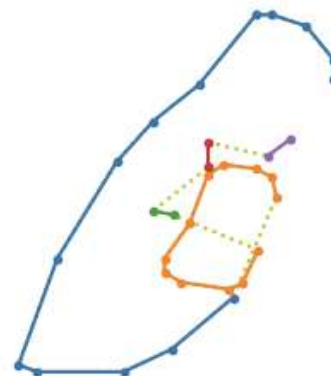
shorts



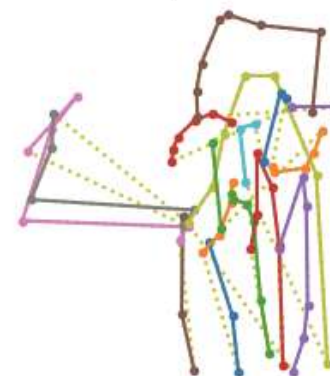
angel



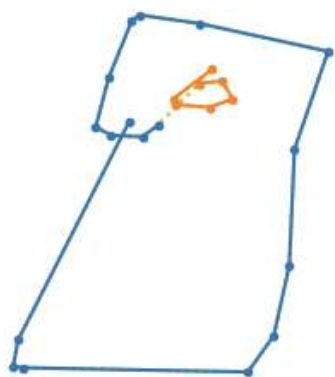
ear



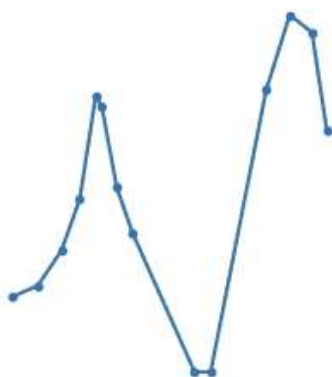
fireplace



remote control



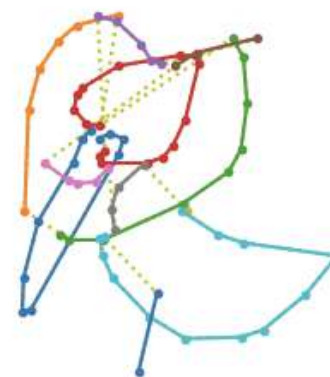
squiggle



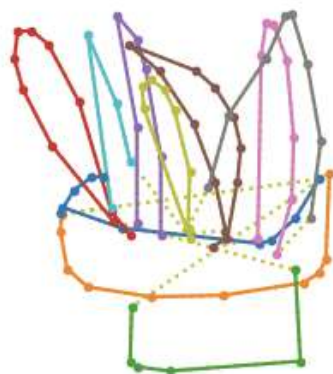
ocean



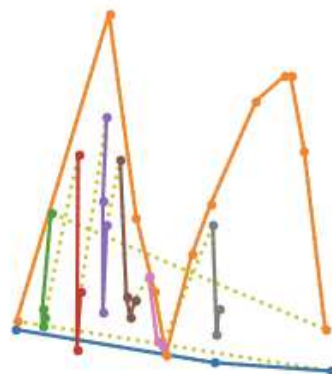
elephant



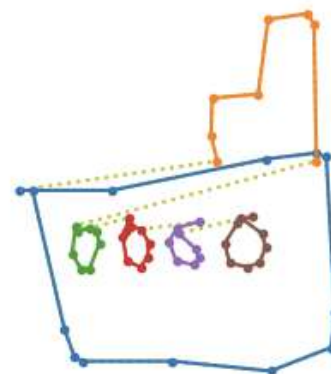
house plant



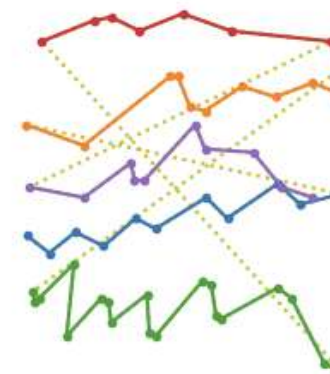
bridge



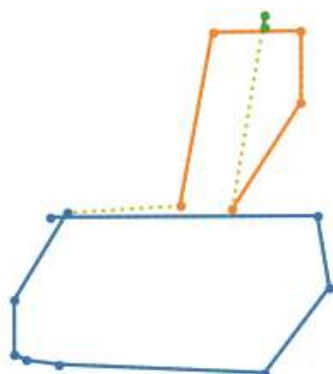
cruise ship



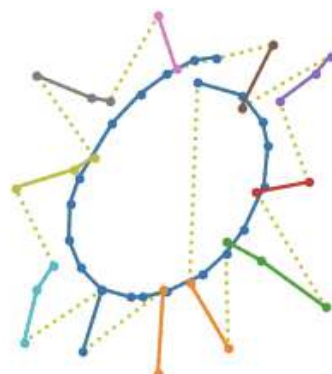
river



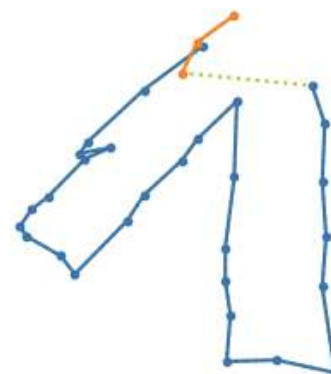
screwdriver



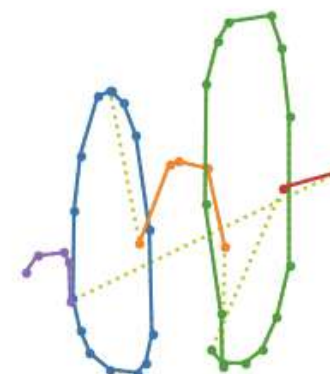
sun



jacket

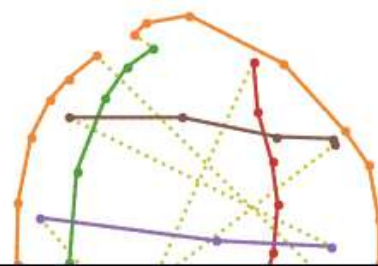


eyeglasses

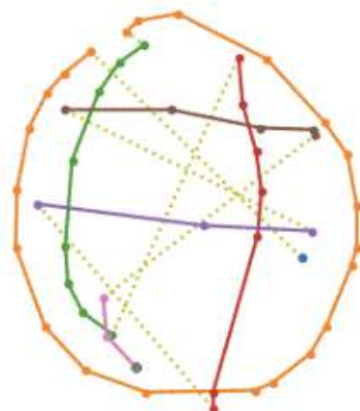


32325/32325 [=====] - 1997s 62ms/step - loss: 5.1678 - accuracy: 0.0403 - sparse_top_k_categorical_accuracy: 0.1301 - val_loss: 4.4608 - val_accuracy: 0.1009 - val_sparse_top_k_categorical_accuracy: 0.2812
5392/5392 [=====] - 149s 27ms/step
1/1 [=====] - 1s 945ms/step

Try to guess



Try to guess

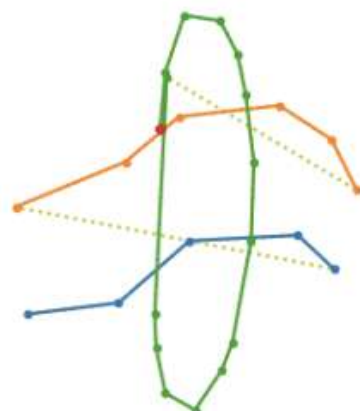


Top-5 predictions:

1. basketball 9.821%
2. onion 4.883%
3. tennis racquet 4.615%
4. baseball 4.122%
5. watermelon 3.986%

Answer: basketball

Try to guess



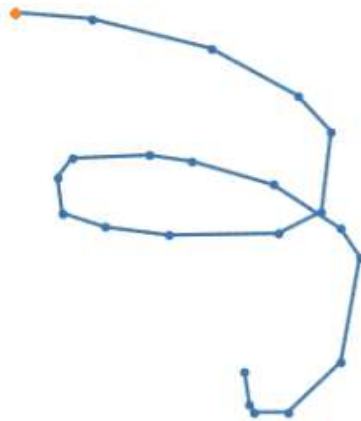
Top-5 predictions:

1. shorts 2.565%
2. pants 2.519%
3. ice cream 2.307%
4. hourglass 2.234%
5. lipstick 2.188%

Answer: belt

Try to guess

Try to guess

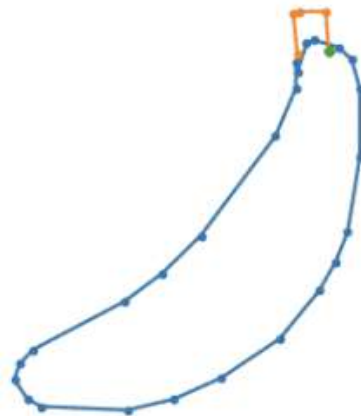


Top-5 predictions:

1. paper clip 5.669%
2. sock 3.088%
3. boomerang 2.503%
4. ear 2.307%
5. shoe 2.251%

Answer: squiggle

Try to guess

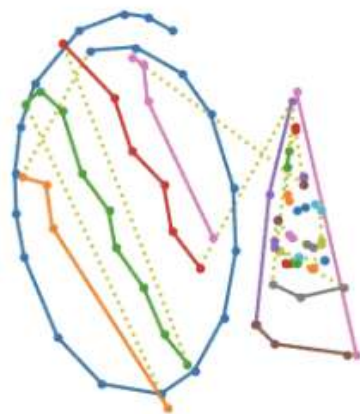


Top-5 predictions:

1. circle 7.783%
2. potato 5.576%
3. pear 4.990%
4. peanut 4.940%
5. moon 4.839%

Answer: banana

Try to guess



Top-5 predictions:
 1. tennis racquet 2.663%
 2. tiger 2.457%
 3. hedgehog 2.414%
 4. brain 2.299%
 5. zebra 2.266%
 Answer: watermelon

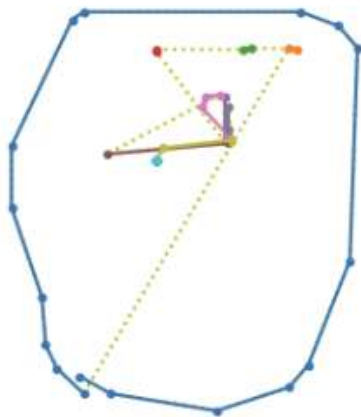
Try to guess



Top-5 predictions:
 1. hand 7.578%
 2. tornado 3.393%
 3. tree 3.310%
 4. saw 3.118%
 5. squiggle 2.862%
 Answer: hand

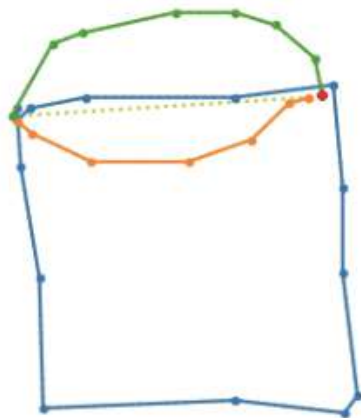
Try to guess

Try to guess



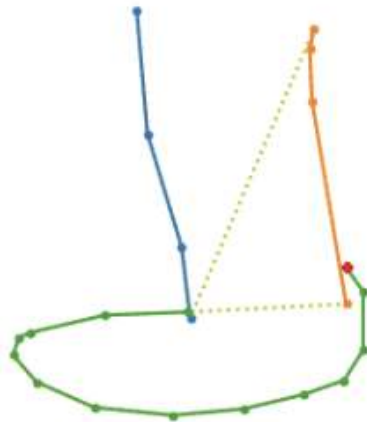
Top-5 predictions:
 1. cake 3.804%
 2. remote control 3.655%
 3. birthday cake 3.356%
 4. map 2.295%
 5. television 2.091%
 Answer: remote control

Try to guess



Top-5 predictions:
 1. picture frame 6.694%
 2. envelope 5.219%
 3. microwave 4.320%
 4. oven 3.808%
 5. television 3.403%
 Answer: bucket

Try to guess

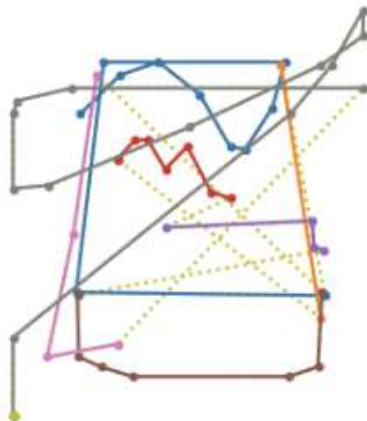


Top-5 predictions:

1. wine bottle 3.064%
2. sock 2.791%
3. baseball bat 2.749%
4. shorts 2.523%
5. boomerang 2.194%

Answer: leg

Try to guess



Top-5 predictions:

1. book 3.568%
2. fence 1.916%
3. sandwich 1.910%
4. couch 1.865%
5. calendar 1.512%

Answer: sock