

Reconnaître un pattern de points de contact sur une surface tactile

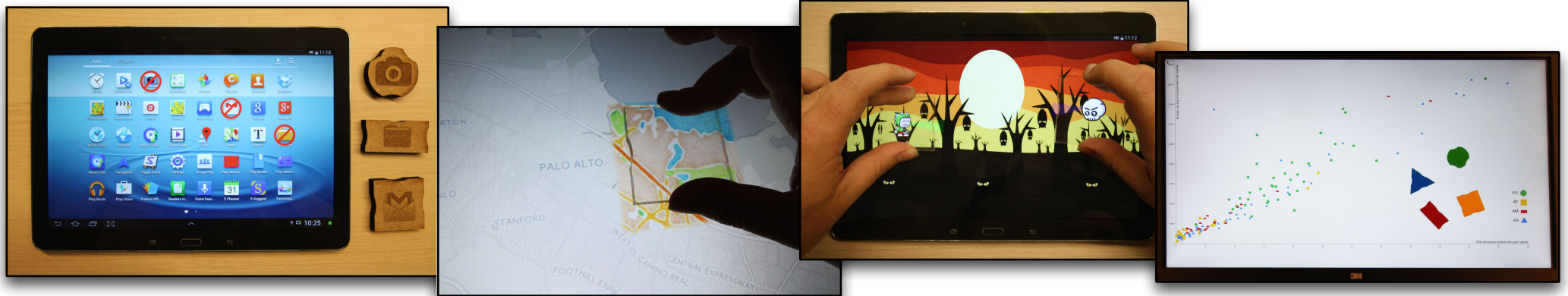
Multi-Touch et Machine Learning

Journée IHM-IA '17
Caroline Appert

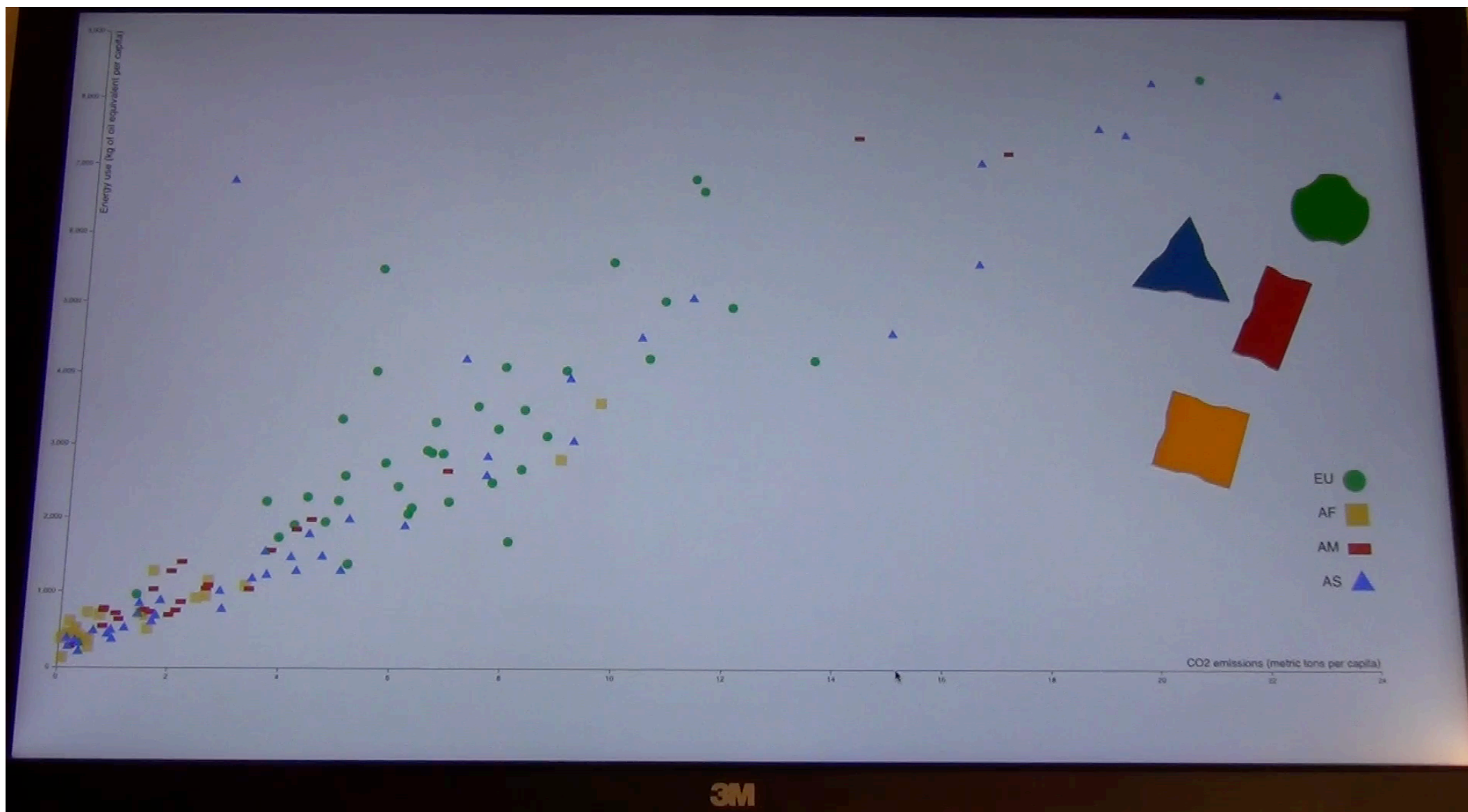


TouchTokens

R. Morales Gonzalez, C. Appert, G. Bailly and E. Pietriga. CHI '16.
TouchTokens: Guiding Multi-Touch Patterns with Passive Tokens.



TouchTokens make it possible
to easily build interfaces that
combine gestures and tangibles



Data visualization



Tangible magic lens



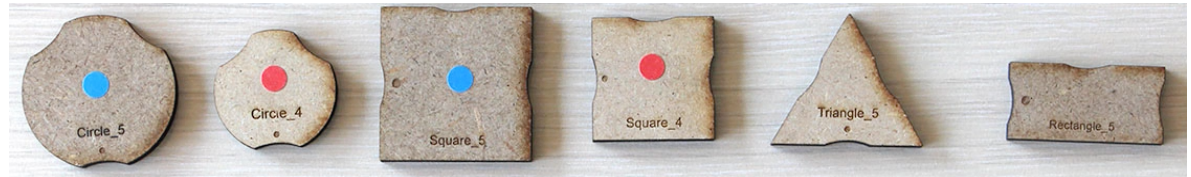
Character controllers in a game



Access control

TouchTokens

Passive tokens indicating
how to grasp them



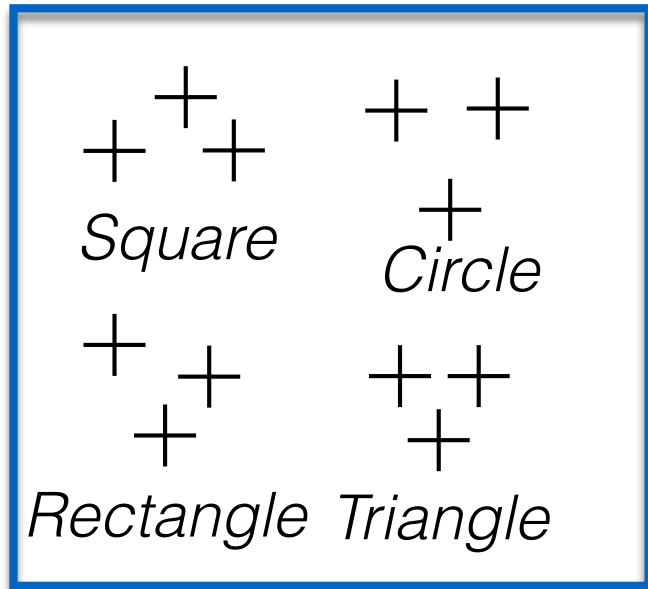
Tactile surface



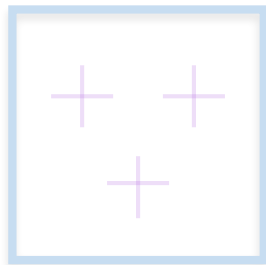
Touch pattern recognizer

Touch pattern recognizer

Library of
template touch patterns



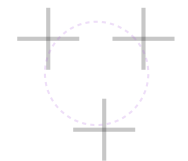
User's input touch
patterns



Recognition
algorithm

Compute
best alignment
between touch
patterns

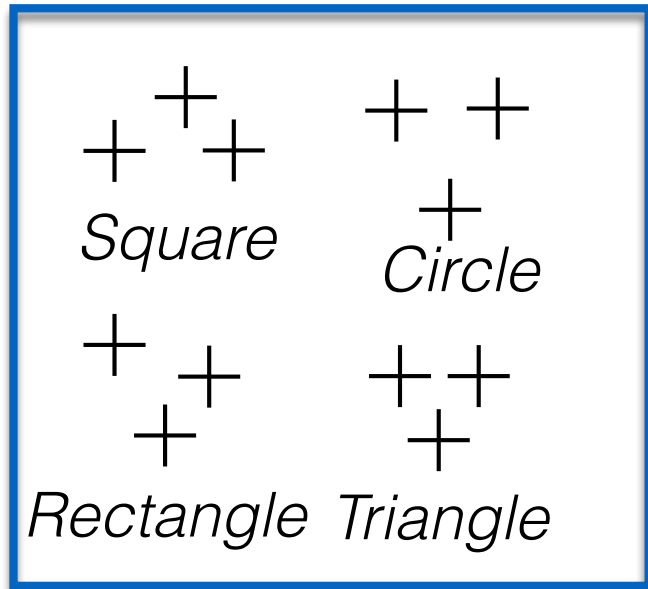
Recognized
token



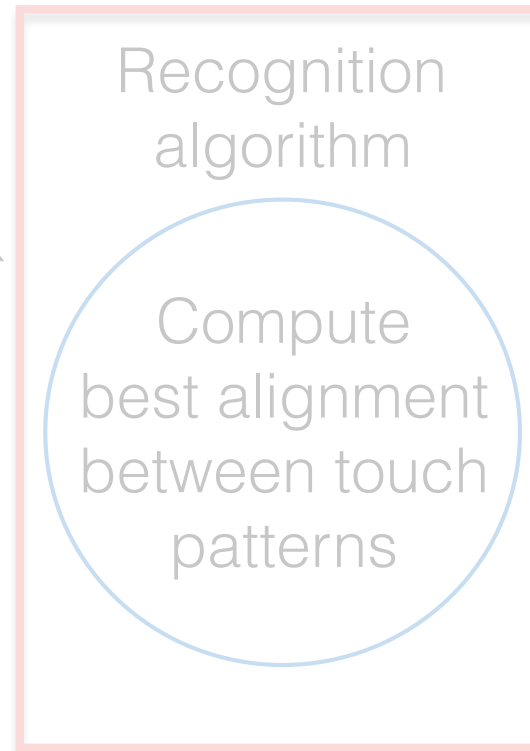
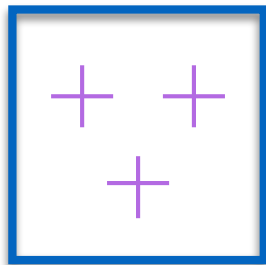
Circle

Touch pattern recognizer

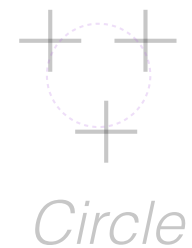
Library of
template touch patterns



User's input touch
patterns

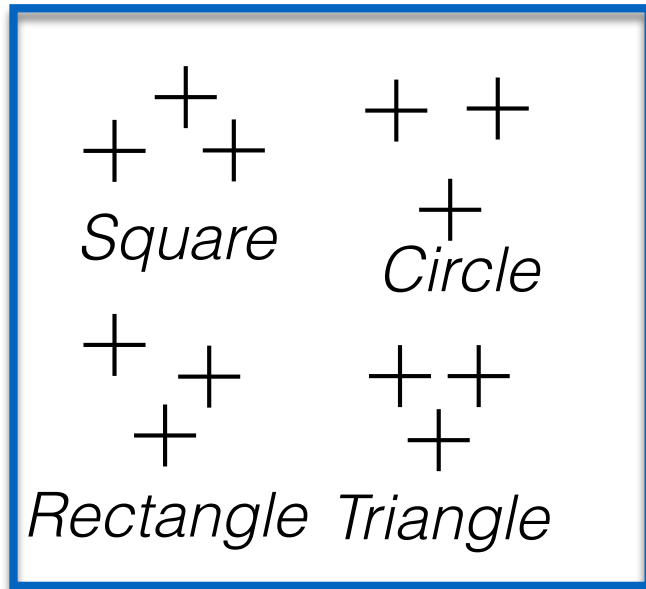


Recognized
token

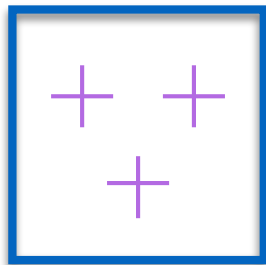


Touch pattern recognizer

Library of
template touch patterns



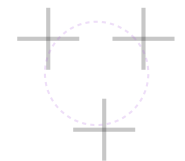
User's input touch
patterns



Recognition
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Compute
best alignment
between touch
patterns

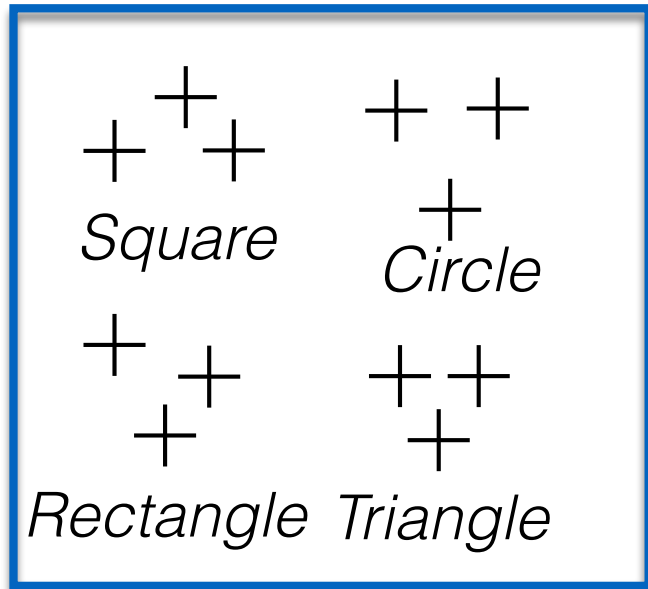
Recognized
token



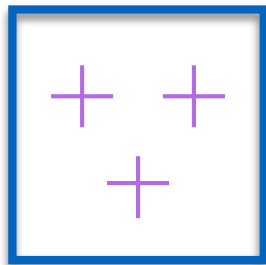
Circle

Touch pattern recognizer

Library of
template touch patterns



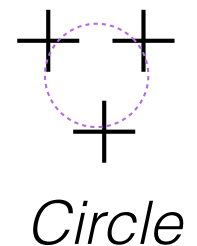
User's input touch
patterns



Recognition
algorithm

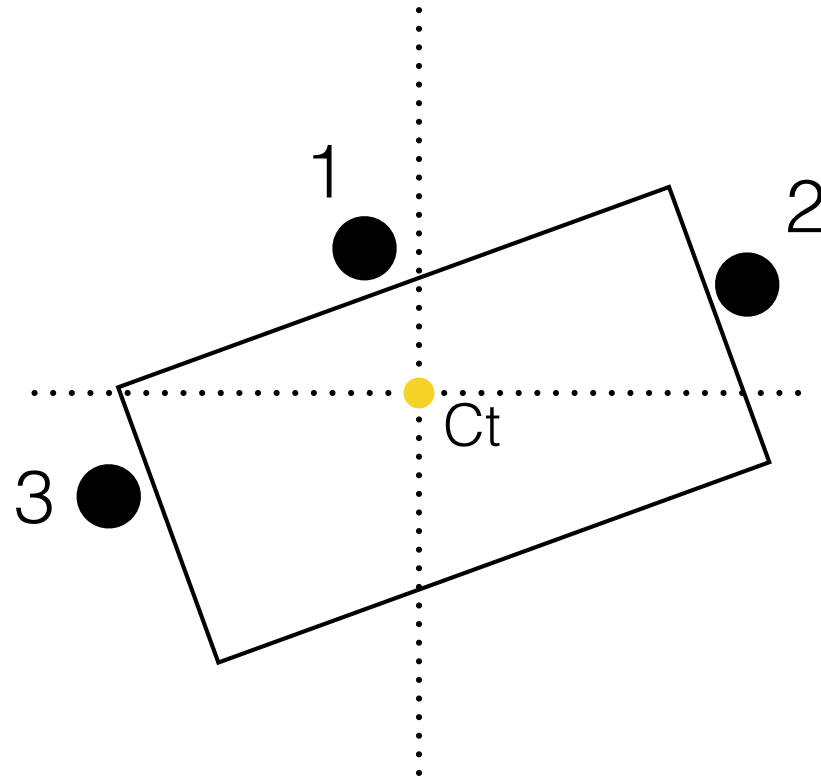
Compute
best alignment
between touch
patterns

Recognized
token



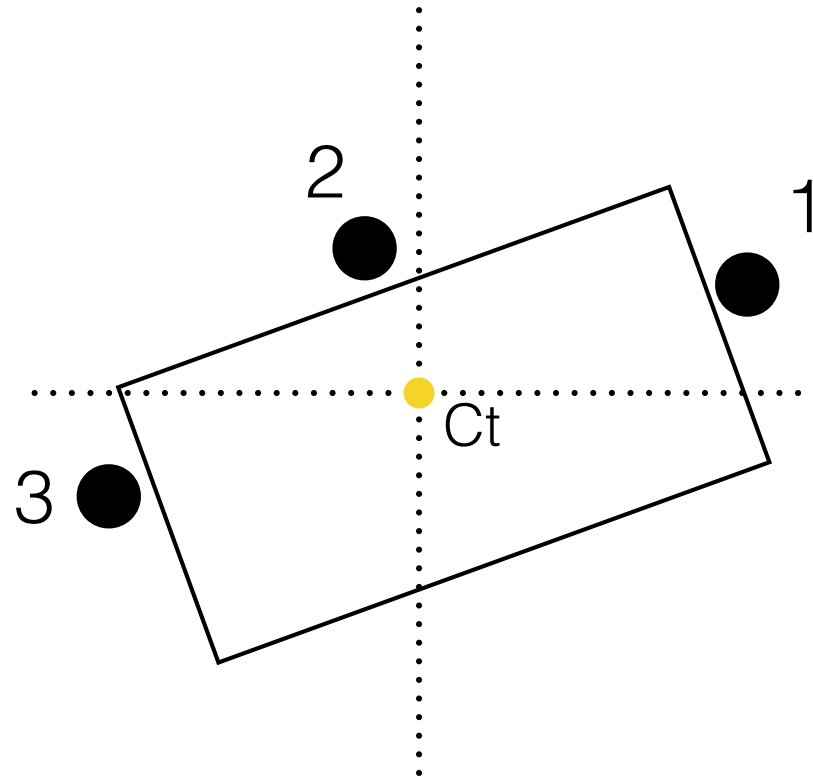
Best alignment algorithm

Template

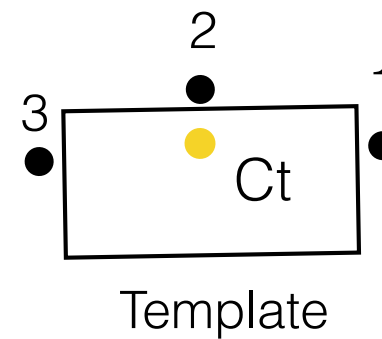
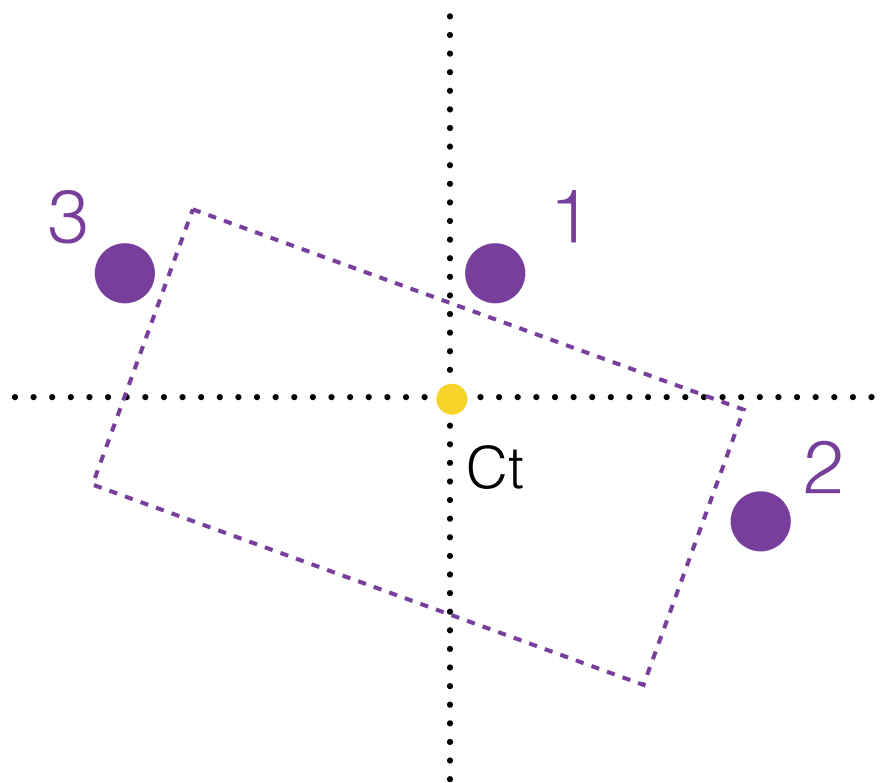


Best alignment algorithm

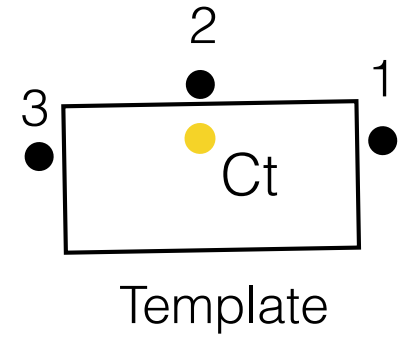
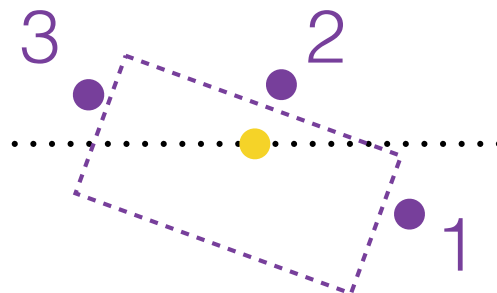
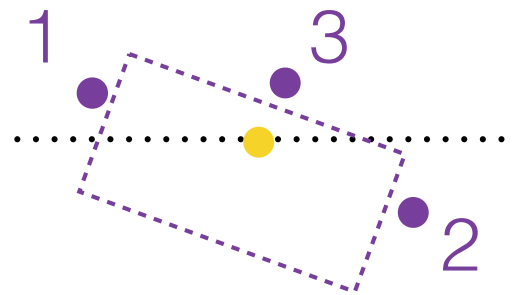
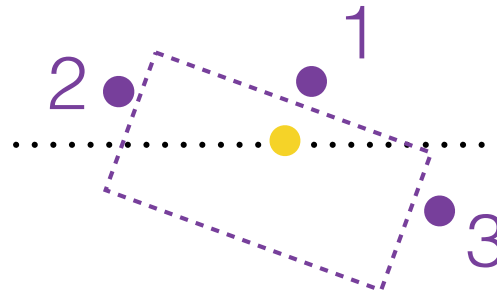
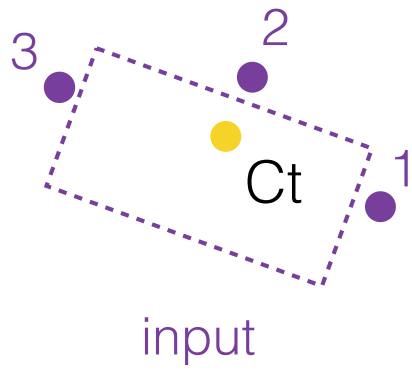
Template



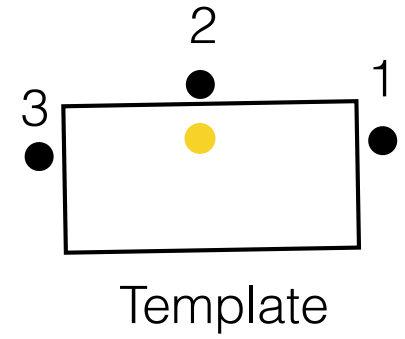
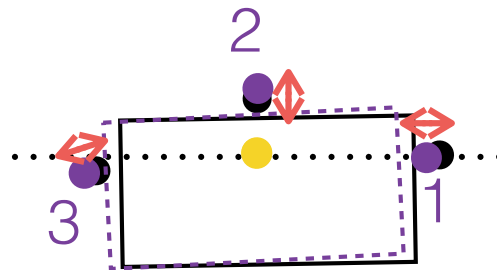
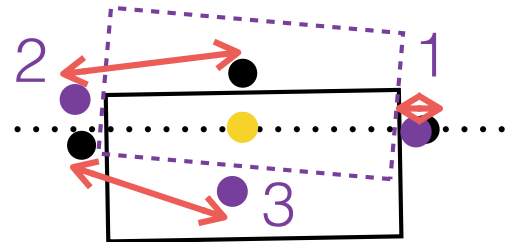
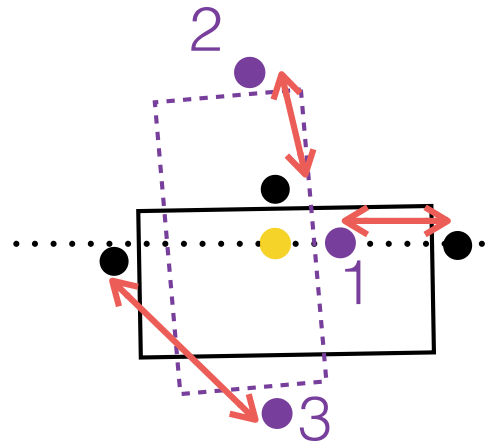
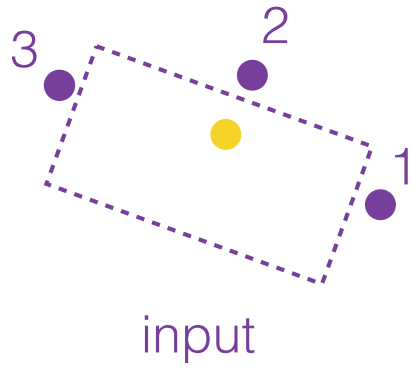
Input



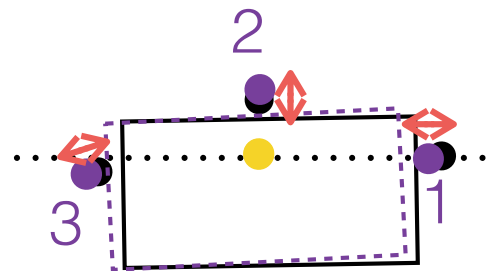
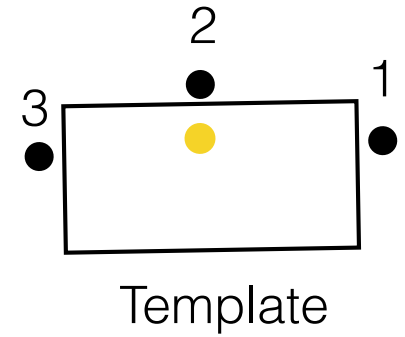
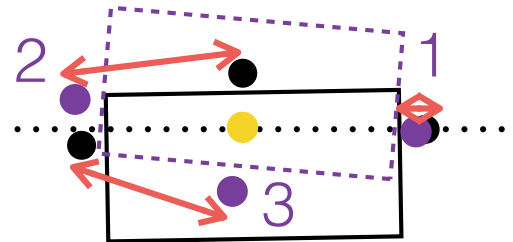
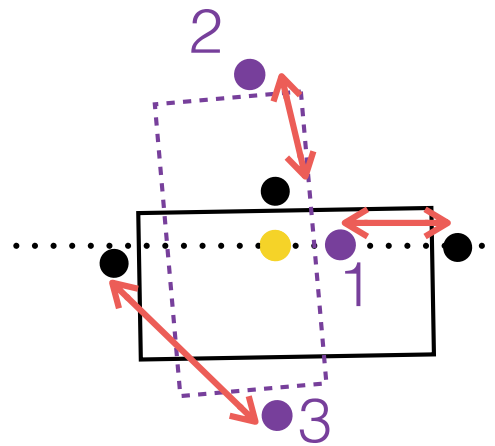
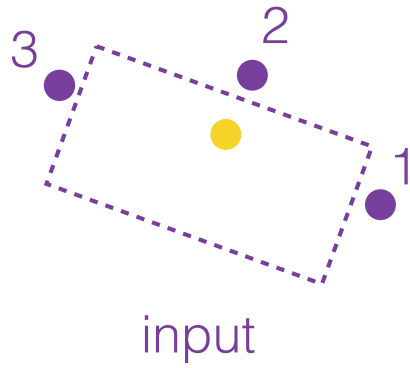
permutations



permutations



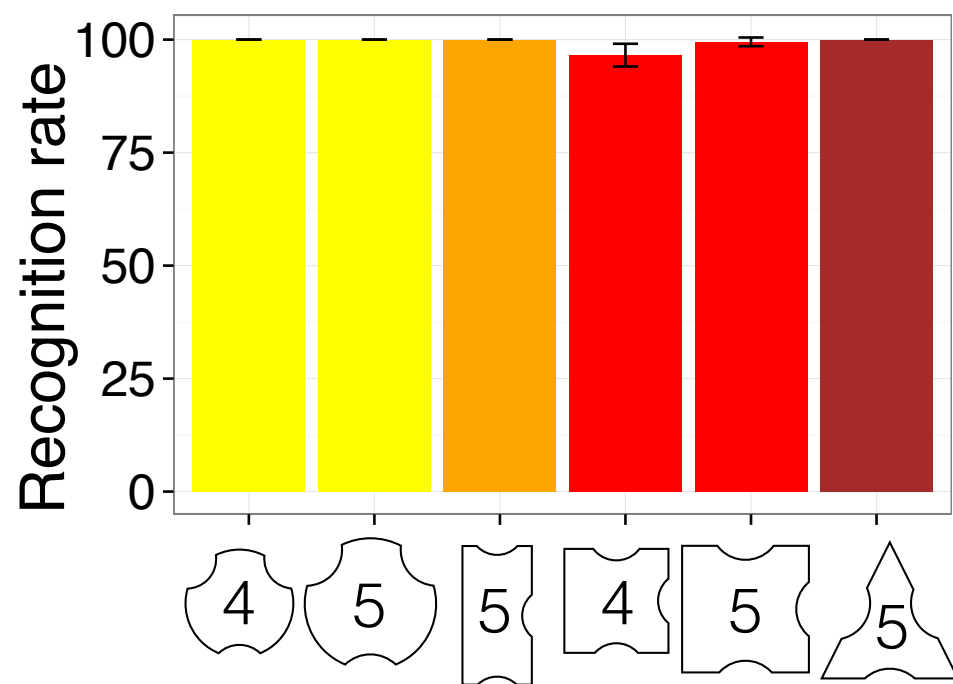
permutations



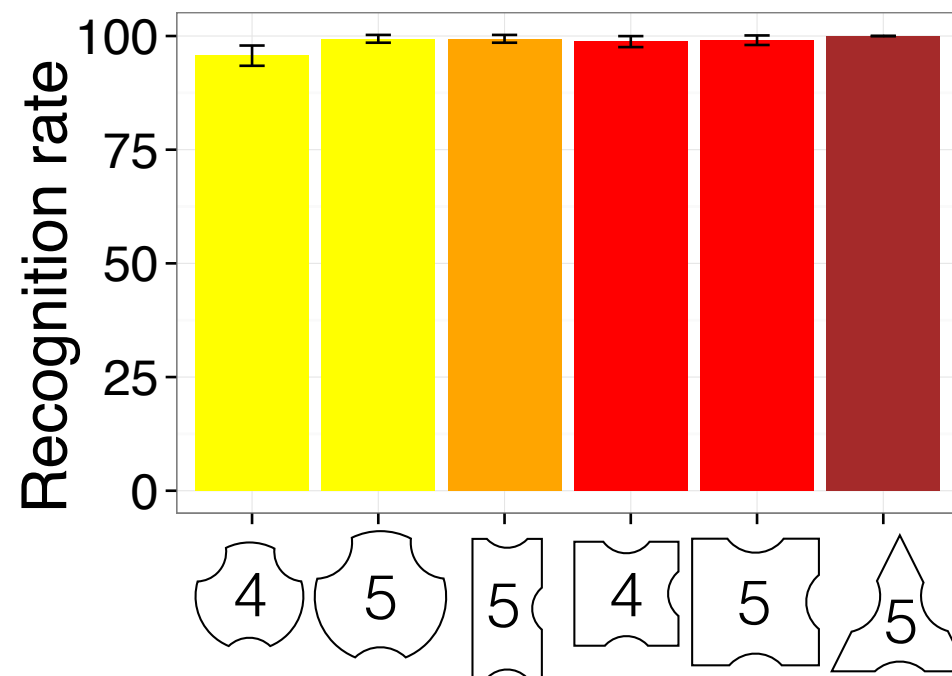
Best Alignment

Results

Tablet: 99.3%

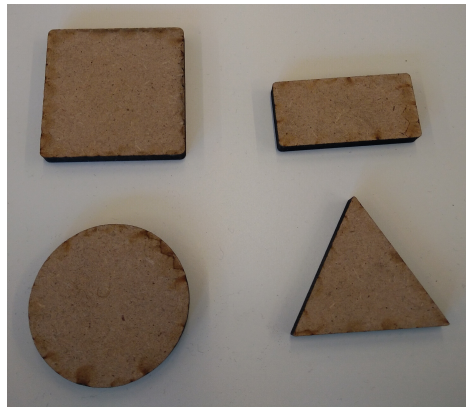


Tabletop: 98.7%



TouchTokens...

...are nice but the initial idea was even nicer!



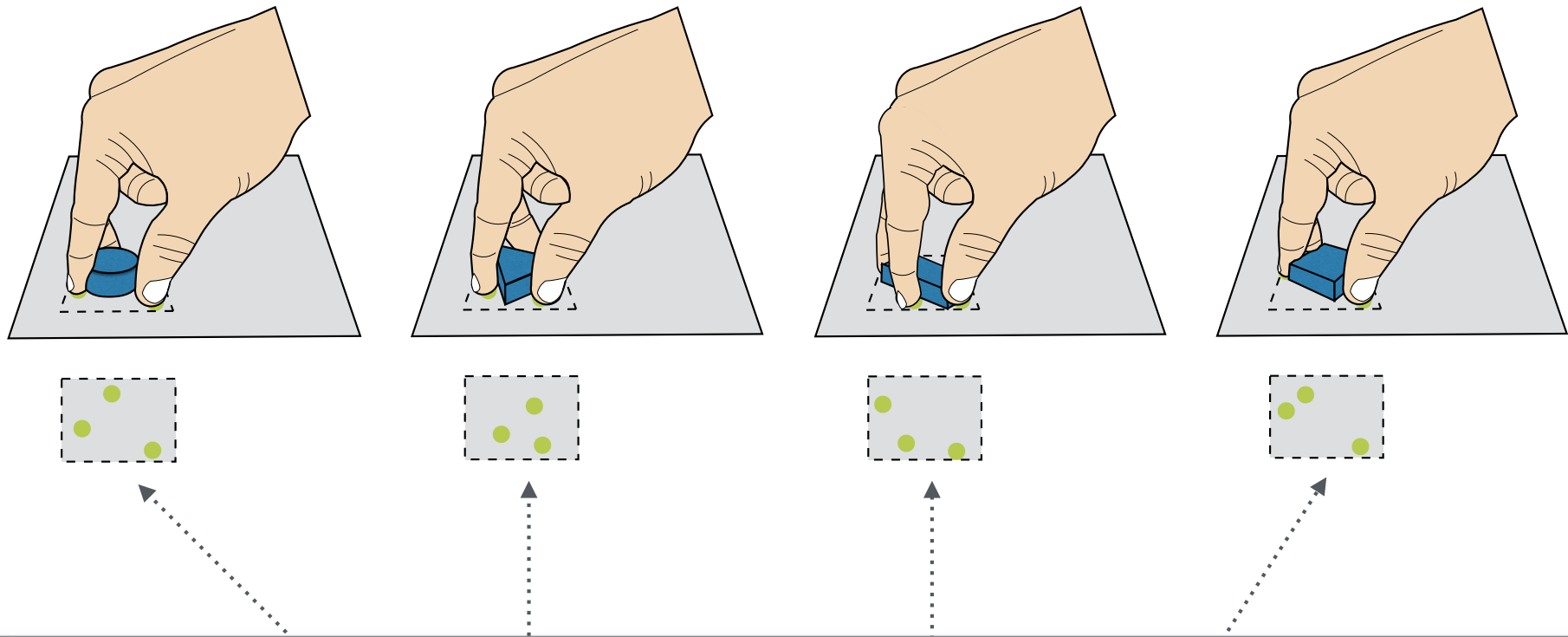
Token without notches



Everyday objects

Let's see where we failed

Initial hypothesis

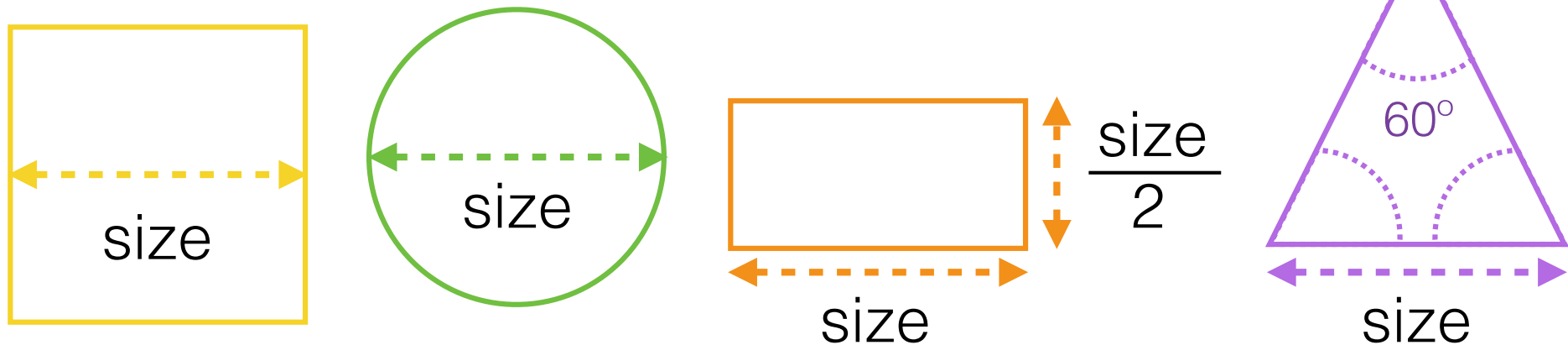


The shape of an object suggests a specific grasp, and thus a specific touch pattern. Objects that significantly differ in shape could thus be discriminated based on their touch pattern.

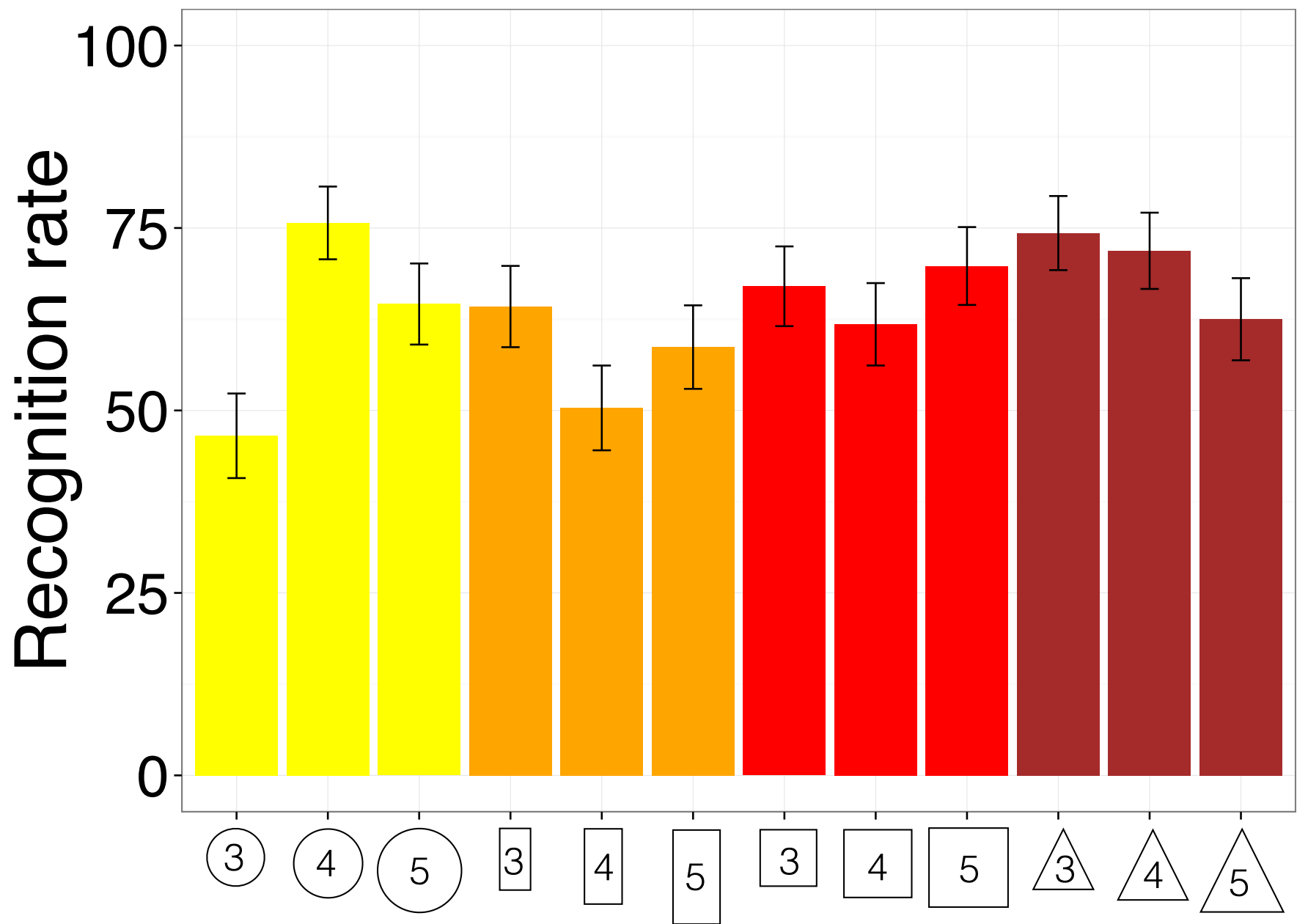
First study

Collection of touch patterns
12 “regular” tokens (no notches)

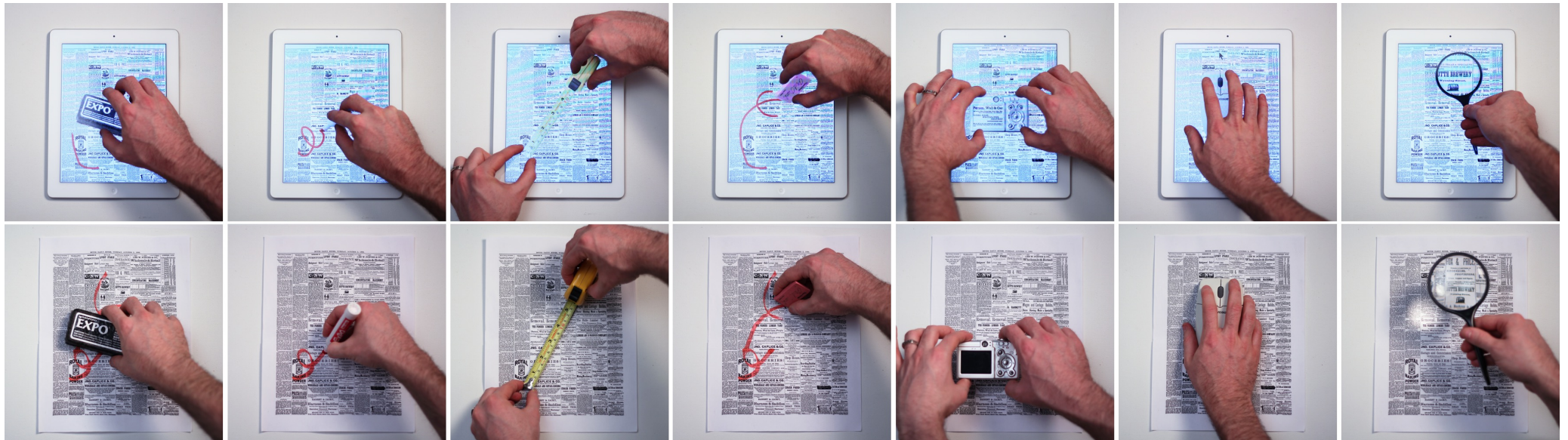
- Size = {3cm, 4cm, 5cm}
- Shape = {Square, Circle, Rectangle, Triangle}
- Size x Shape = 12 geometries



Results



Could ML help?



TouchTools

Chris Harrison, Robert Xiao, Julia Schwarz, and Scott E. Hudson. CHI '14. TouchTools: leveraging familiarity and skill with physical tools to augment touch interaction.

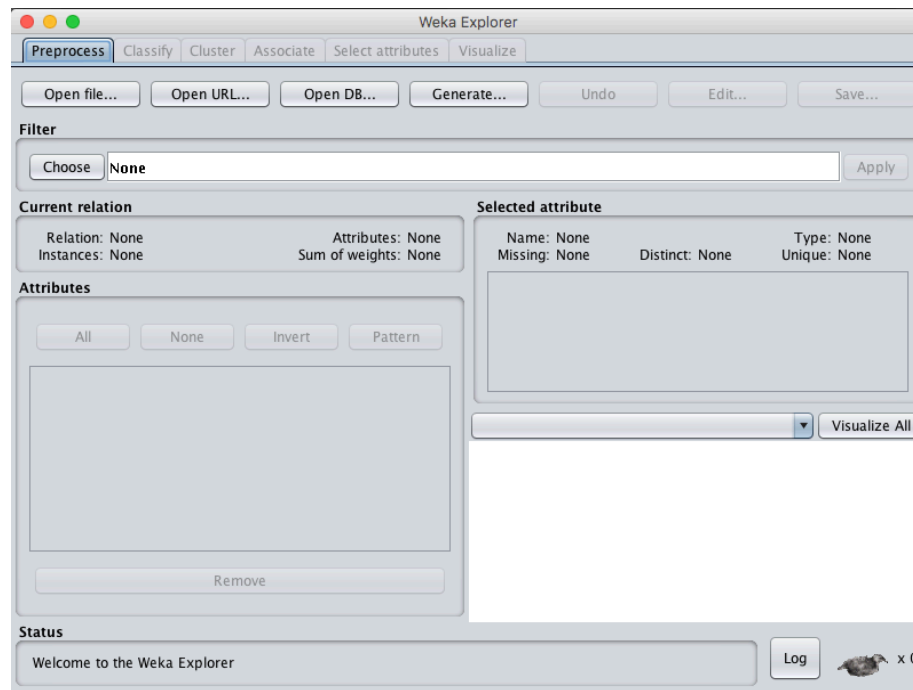
147 training instances in total (7 participants x 7 tools x 3 repetitions)

[...]

Our **features** are as follows: the number of touch points, the estimated total touch area, [...], angles between consecutively-clockwise points as measured from the centroid, and the size of each point. These features are then fed into a **support vector machine classifier** trained on previously recorded data.

Let's give it a try

scikit-learn? Weka?



Weka explorer: a GUI! Nice!

Feed Weka explorer

What features?

Let's try all the features we can think of

features.arff

```
@relation whatever
```

```
@attribute Participant numeric
```

```
@attribute FingersCount numeric
```

```
@attribute TouchArea numeric
```

```
@attribute SmallestInscribedCircleRadius numeric
```

```
@attribute MeanPairedDistance numeric
```

```
@attribute MedianPairedDistance numeric
```

```
@attribute MinPairedDistance numeric
```

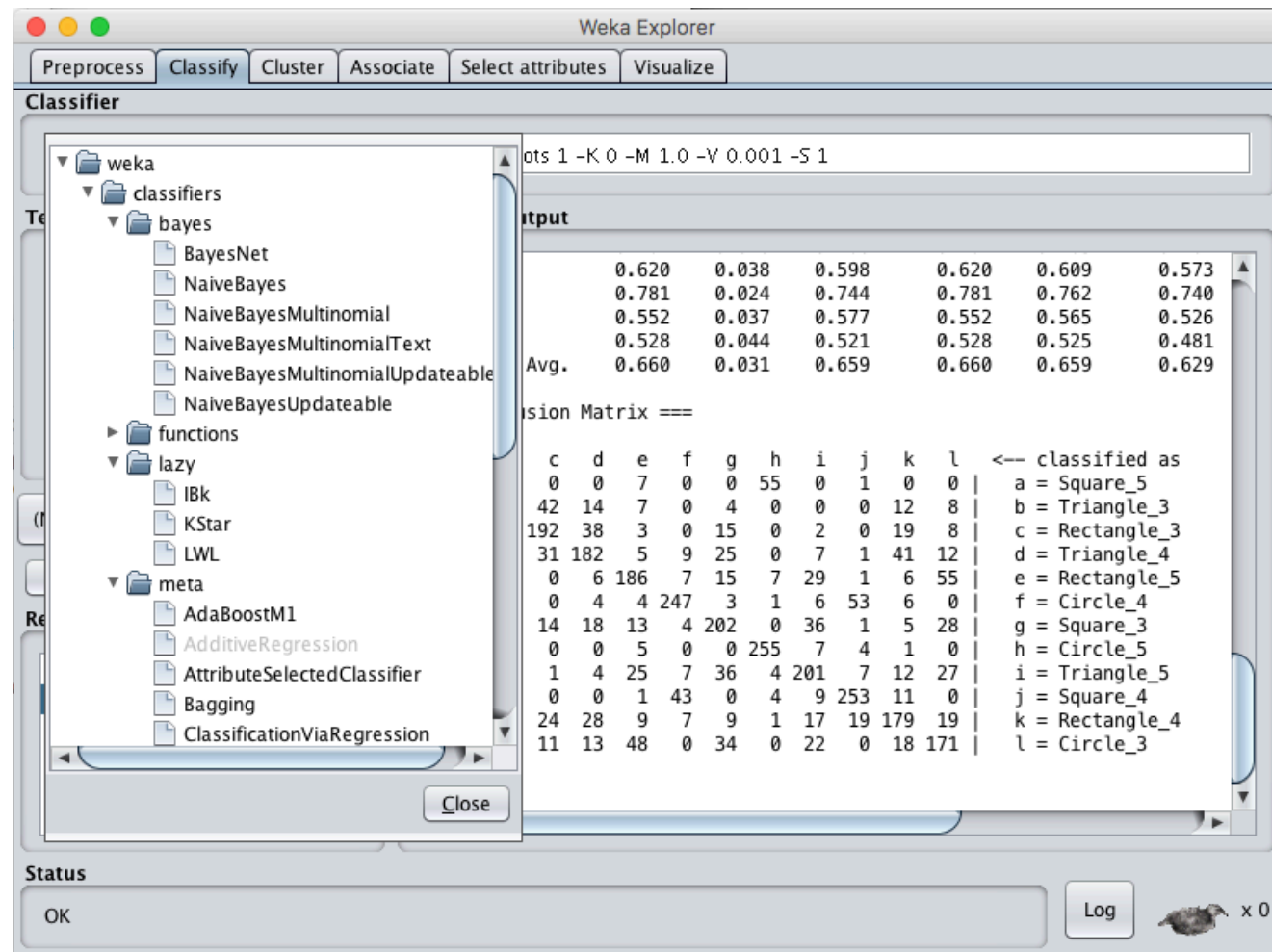
```
@attribute MaxPairedDistance numeric
```

```
[...]
```

```
@data
```

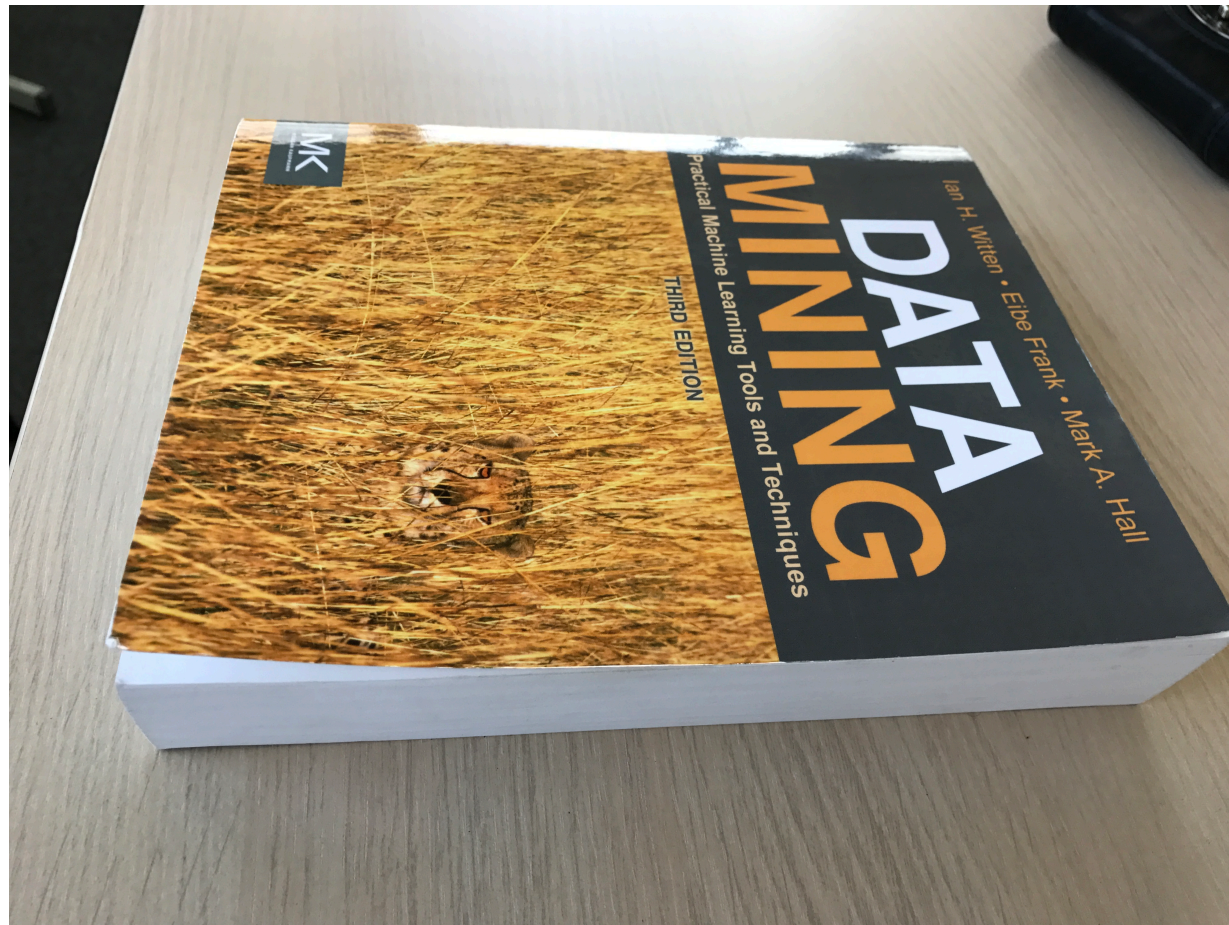
```
2, 4, -9850, 80.42, 120.68, 127.97, 46.69, 163.67, 43.31, 75.95, 69.7, 60.37, 104.03,  
19.29, 81.17, 80.42, 80.42, 83.43, 1.5, 1.57, 1.67, 0.7, 2.24, 0.64, Square_5  
2, 3, -1605, 36.8, 62.34, 64.4, 51.79, 70.84, 9.69, 36.02, 35.36, 30.99, 41.7, 5.39,  
36.8, 36.8, 36.8, 36.8, 0, 2.09, 2.17, 1.79, 2.33, 0.28, Triangle_3
```

Explore with Weka Explorer



So many concepts to understand...

Learn Machine Learning



It takes (way) too much time...

In the end...

Alternative approaches we considered led to significantly poorer performance. In particular, we tested k-Nearest-Neighbour (k=1 and k=3) and SVM algorithms, using both raw data and describing features. The raw data was pre-processed to make it independent from rotation angle and finger identification. The describing features we considered included the touch envelope's area, as well as various descriptive statistics (min, max, mean, median and standard deviation) for measures such as point-centroid distance, distance between successive points, distance between any pair of points, etc. **These machine learning approaches yielded recognition rates ranging from 50% to 85% per participant.**

Why?

Is ML not able to help?

Is the problem unsolvable?

Is the problem not captured in the right way (i.e., wrong describing features)?

Did we just fail to use it properly?

The design space for ML-based recognizers is very large: type of classifier, feature selection, size of training set, ...

Why?

Is ML not able to help?

Is the problem unsolvable?

Is the problem not captured in the right way (i.e. wrong describing features)?

We could probably answer
this with good tools

Did we just fail to use it properly?

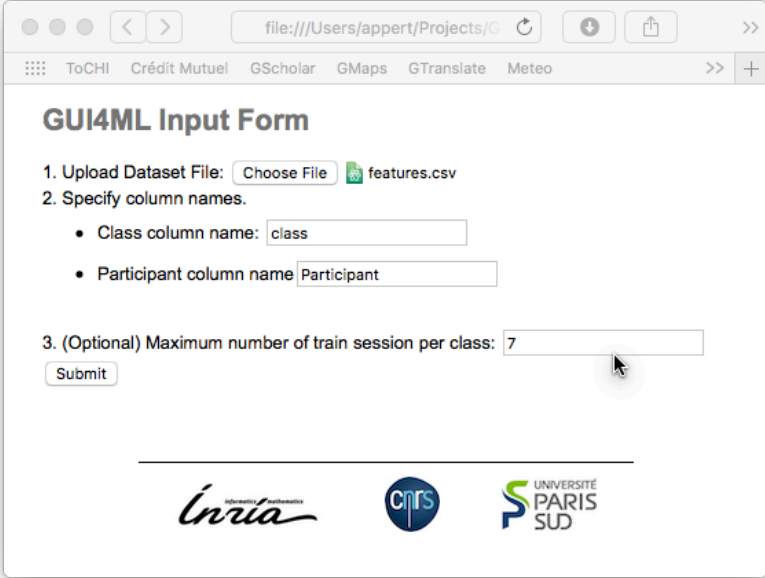
The design space for ML-based recognizers is very large: type of classifier, feature selection, size of training set, ...

What would be nice for designing MT recognizers

Collect empirical data

Compute describing features

Load it in a very simple GUI...



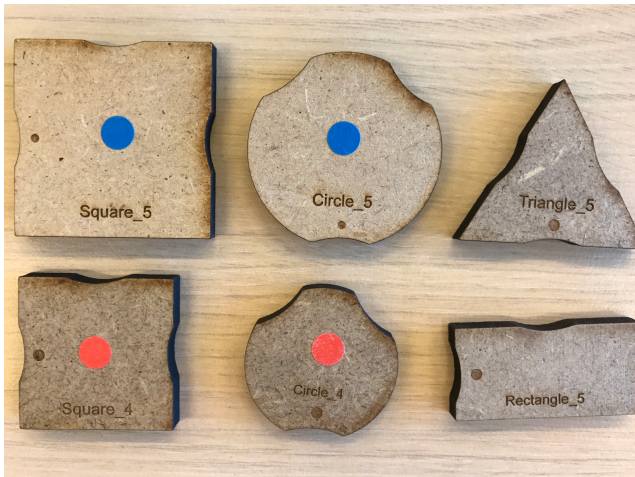
The screenshot shows a web browser window displaying the 'GUI4ML Input Form'. The browser's address bar shows a local file path: 'file:///Users/appert/Projects/G...'. The browser's tab bar includes several tabs: 'ToCHI', 'Crédit Mutuel', 'GScholar', 'GMaps', 'GTranslate', and 'Meteo'. The form itself has a title 'GUI4ML Input Form' and three main sections:

- 1. Upload Dataset File:** A 'Choose File' button and a file icon next to the text 'features.csv'.
- 2. Specify column names:**
 - 'Class column name:' followed by a text input field containing 'class'.
 - 'Participant column name:' followed by a text input field containing 'Participant'.
- 3. (Optional) Maximum number of train session per class:** A text input field containing the number '7'.

Below the third section is a 'Submit' button. At the bottom of the form, there is a horizontal line separating it from the footer, which contains three logos: 'Inria', 'CNRS', and 'UNIVERSITÉ PARIS SUD'.

...that systematically explores the design space of ML-based recognizers and tells me what the best solution is with different training strategies (global vs per-user)

Multi-touch recognizer and training



Pre-defined interfaces should ideally rely on a global training strategy

train with n users and test with the $(n+1)^{\text{th}}$ user



Customizable interfaces require a per-user training strategy

train with $(x - y)$ examples from user n
and test with the y remaining examples
from user n

Multi-Touch & Machine Learning

Take-away message

We need easy-to-use tools that systematically explore the design space of ML-based recognition engines

Open question

Could ML help in designing incremental recognition engines for gestures with a continuous effect (e.g., pinch-to-zoom)?

Multi-Touch & Machine Learning

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