

Visual measurement of jackets by structured prediction

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Problem

A company sells jackets custom tailored to the sizes specified by their clients through its website.

The screenshot shows a website for 'CUSTOM CLOTHING MADE EASIER'. The main banner features a man in a dark suit with a red and blue striped tie. Text on the banner includes 'CUSTOM SUITS FROM 189€ DESIGN NOW >', 'CUSTOM SHIRTS FROM 42€ DESIGN NOW >', and 'Linen Waves 100% Linen Shirts Collection'. Below the banner, there's a section titled 'LATEST NEWS' with three cards: 'GET A FREE SUIT' (with a cartoon character), 'Linen Waves 100% Linen Shirts Collection' (with a man in sunglasses), and 'Discover our newest clothing line for women' (with a woman in a blazer). Each news card has a 'See More' button at the bottom.

Problem

Sometimes customers return items complaining they don't fit well.

Causes :

- ▶ customers don't measure themselves well with a tailor tape
- ▶ input wrong measures to the web page
- ▶ tailor had a bad day
- ▶ ??

There's a need for comparing the actual measures of an item with those provided by customers.

Goal

Perform automatic measurements on each garment at *pre-shipment time* for quality control, by means of computer vision techniques.

Problem

What to measure



Image capture

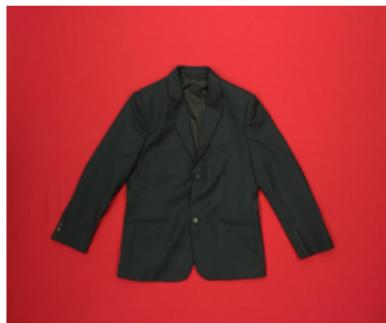


Samples

We captured a first set of 23 jackets as working samples, from which

- ▶ 16 normal color, style, size
- ▶ 7 less frequent

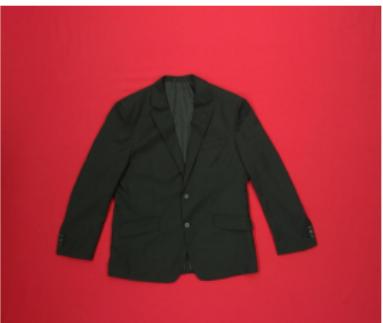
Normal samples



1



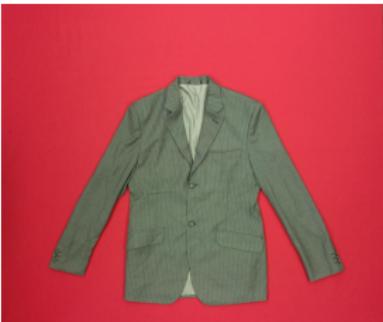
2



3



4

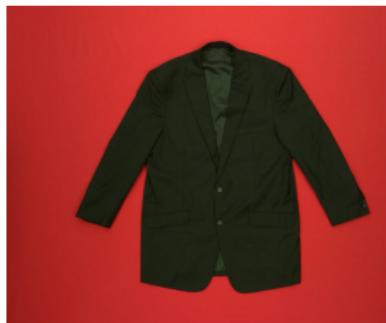


5

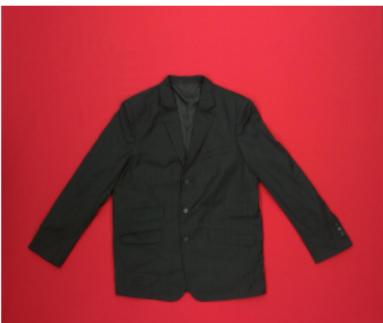


6

Normal samples



7



8



9



10



11

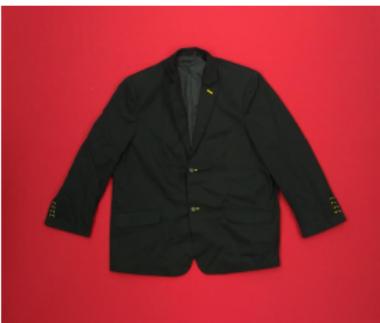


12

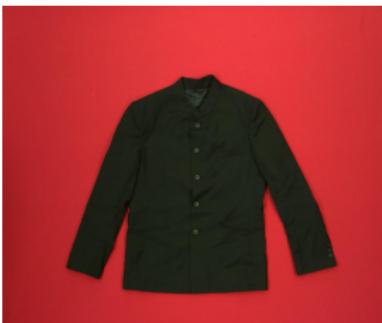
Normal samples



13



14



15



16

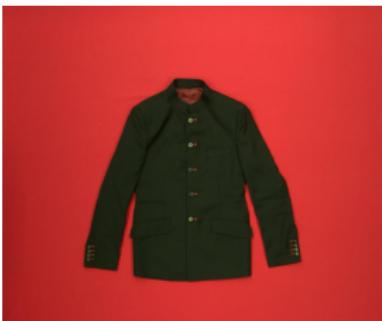
Less frequent samples



17



18



19



20



21



22

Less frequent samples



23

Later we received a second set of 82 more images for which we made an additional groundtruth file.

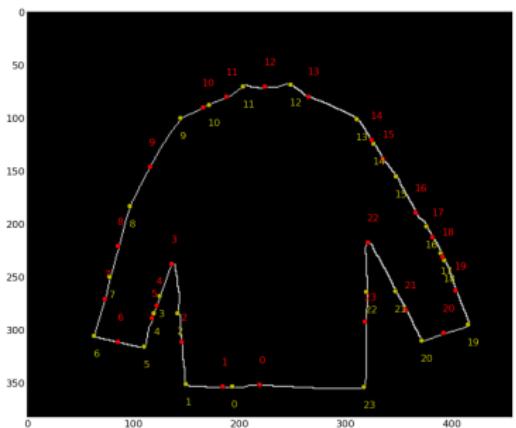
Method

1. Segmentation
2. Extract contour

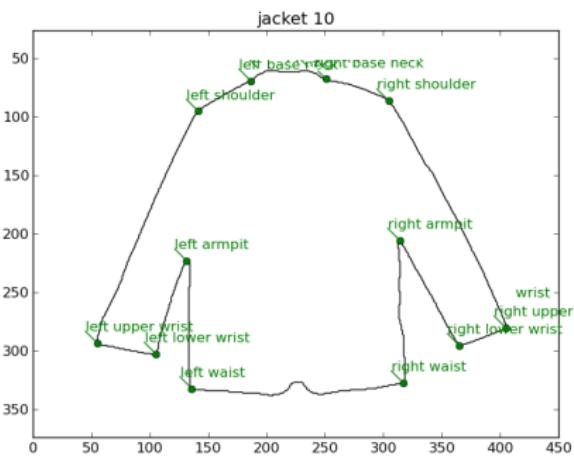


Method

3. contour → curvature → extrema points → keypoints



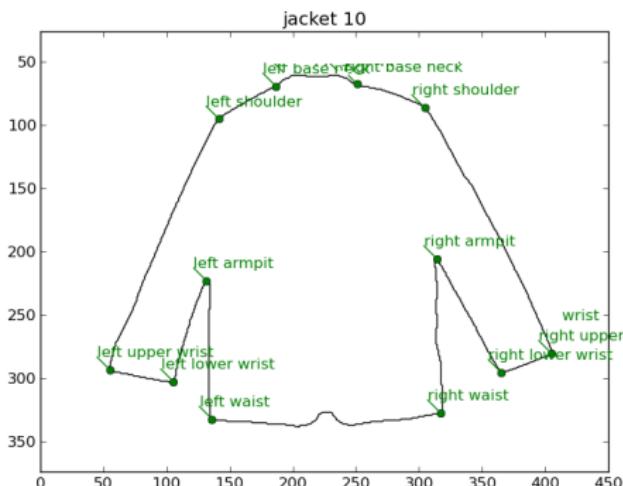
yellow = maxima / convex
red = minima / concave



Method

3. contour → curvature → extrema points → keypoints

```
labels_extrema = {  
    0: 'not a keypoint',  
    1: 'left base neck',  
    2: 'left shoulder',  
    3: 'left upper wrist',  
    4: 'left lower wrist',  
    5: 'left armpit',  
    6: 'left waist',  
    7: 'right waist',  
    8: 'right armpit',  
    9: 'right lower wrist',  
    10: 'right upper wrist',  
    11: 'right shoulder',  
    12: 'right base neck'  
}
```

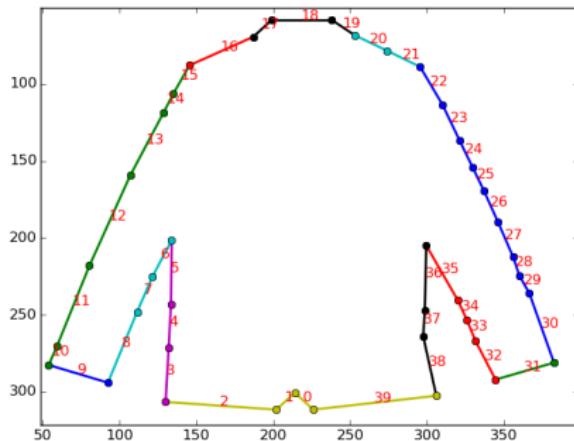


Method

or

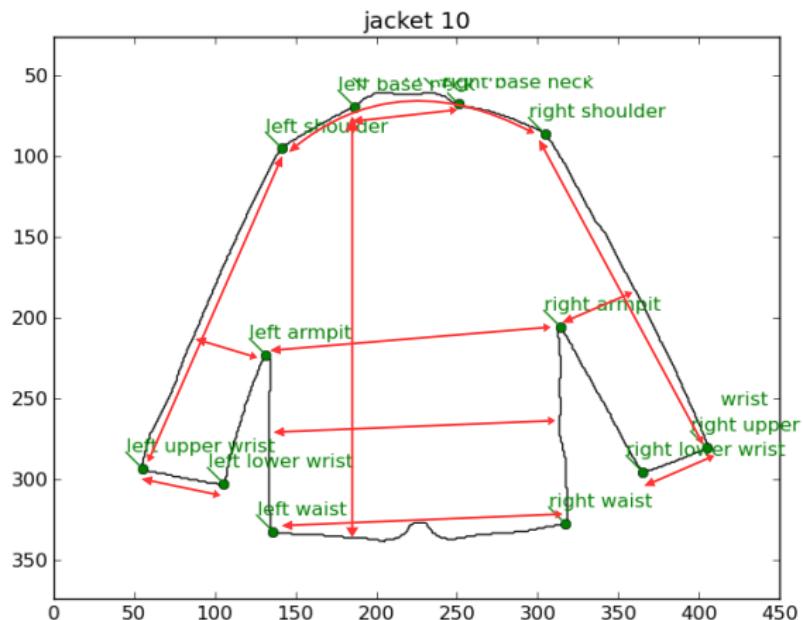
3. contour → curvature → extrema points → segments → keypoints

```
labels_segments = {  
    0: 'neck',  
    1: 'left shoulder',  
    2: 'outer left sleeve',  
    3: 'left wrist',  
    4: 'inner left sleeve',  
    5: 'left chest',  
    6: 'waist',  
    7: 'right chest',  
    8: 'inner right sleeve',  
    9: 'right wrist',  
    10: 'outer right sleeve',  
    11: 'right shoulder',  
}
```



Method

4. keypoints → measures



Method

What should we label = classify to then get the keypoints ?

- ▶ curvature extrema, or
- ▶ segments between curvature extrema

What has a more constraining context ? (see the labels in each case)

What to do

Given

- ▶ segments plus some precomputed features
- ▶ groundtruth = type of each segment

for each jacket, **learn a segment classifier** (no need to obtain keypoints, it's direct)

What to do

1. design a **graphical model** to infer the class of each segment, making the most of the structured nature of the problem
2. train it = **learn** its parameters, trying a few different ssvm learning algorithms
3. perform structured prediction as **inference** on it
4. **evaluation** with kfold=5 on jackets (23 for learning, 4 for testing)
5. compare performance with classification by standard linear SVM
6. **explain** results

Tools, data & code

- ▶ Python + some IDE (for example Windows: Python(x, y),
Windows/Linux: Anaconda)
- ▶ scikit-learn for linear SVM and cross-validation
- ▶ Panda or cvs to load Excel spreadsheets
- ▶ Pystruct + dependencies (CVXOPT, qpbo)
- ▶ Pystruct OCR letter recognition example !
- ▶ segments for 23 jackets, groundtruth and template script

Grading

- ▶ C = 0 points : code does not work or performs bad, doesn't learn a GM
- ▶ B = 0.5 points : it works and performs well (like >95% accuracy) with the first set of 23 samples + comparison with linear SVM + good discussion of results
- ▶ A = 1 point : B + add second set of 82 samples + experiment with different sets of features. Caution, new images were not recorded by us, some are problematic.

Deliverables

Each group sends to joans@cvc.uab.es

- ▶ the source code, that can or not be based on the template we provide, plus 10-15 slides in PDF or so, with
- ▶ quantitative + qualitative results
- ▶ discuss results : when does it work ? when not ? comparison
- ▶ all in a single zip file

Include names of group members and NIU in the email text.