

## Descriptions Using Moments

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## Moments

Family of stable binary (and grey level) shape descriptions

Can be made invariant to translation, rotation,

Let  $\{p_{rc}\}$  be the binary (0,1) image pixels for row  $r$  and col  $c$  where 1 pixels are the object

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## Moments II

Area  $A = \sum_r \sum_c p_{rc}$

Center of mass

$(\hat{r}, \hat{c}) = (\frac{1}{A} \sum_r \sum_c r p_{rc}, \frac{1}{A} \sum_r \sum_c c p_{rc})$

A family of 'central' () invariant moments (for any  $u$  and  $v$ ):

$$m_{uv} = \sum_r \sum_c (r - \hat{r})^u (c - \hat{c})^v p_{rc}$$

Subtracting center of mass makes it translation invariant

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## Scale invariant moments

If double in dimensions, then moment  $m_{uv}$  increases by  $2^u 2^v$  for weightings and 4 for the number of pixels.

Similarly, area  $A$  increases by 4, and thus  $A^{(u+v)/2+1}$  increases by  $4 \times 2^u 2^v$

So, the ratio:

$$\mu_{uv} = \frac{m_{uv}}{A^{(u+v)/2+1}}$$

is invariant to .

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## Rotation invariant moments

Moment invariant theory has identified methods to generate various orders of moments invariant to rotation.

6 functions  $ci_i$  with rescaling applied to get into similar  ranges

Area  $A = \sum_r \sum_c p_{rc}$   
Center of mass  $(\hat{r}, \hat{c})$

Define complex  $uv$  central moment:

$$c_{uv} = \sum_r \sum_c ((r - \hat{r}) + i(c - \hat{c}))^u ((r - \hat{r}) - i(c - \hat{c}))^v p_{rc}$$

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invariance

Get specific scale invariant moments:

$$s_{11} = c_{11}/(A^2)$$

$$s_{20} = c_{20}/(A^2)$$

$$s_{21} = c_{21}/(A^{2.5})$$

$$s_{12} = c_{12}/(A^{2.5})$$

$$s_{30} = c_{30}/(A^{2.5})$$

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## Rotation invariant moments II

Rescaled (so values in similar range) rotation invariants:

$$ci_1 = \text{real}(s_{11})$$

$$ci_2 = \text{real}(1000 * s_{21} * s_{12})$$

$$ci_3 = 10000 * \text{real}(s_{20} * s_{12} * s_{12})$$

$$ci_4 = 10000 * \text{imag}(s_{20} * s_{12} * s_{12})$$

$$ci_5 = 1000000 * \text{real}(s_{30} * s_{12} * s_{12} * s_{12})$$

$$ci_6 = 1000000 * \text{imag}(s_{30} * s_{12} * s_{12} * s_{12})$$

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Moment matlab code

```
function vec = getproperties(Image)
    area = bwarea(Image);
    perim = bwarea(bwperim(Image,4));
    compactness = perim*perim/(4*pi*area);
    s11 = complexmoment(Image,1,1) / (area^2);
    s20 = complexmoment(Image,2,0) / (area^2);
    ...
    ci1 = real(s11);
    ci2 = real(1000*s21*s12);
    ci3 = 10000*real(s20*s12*s12);
    ...
```

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## Example property values



compactness	1.93	1.81	1.90
$ci_1$	0.23	0.27	0.25
$ci_2$	0.18	0.37	0.45
$ci_3$	0.08	-0.50	0.11
$ci_4$	-0.00	0.37	-0.64
$ci_5$	0.23	-0.47	0.09
$ci_6$	-0.00	0.07	-0.63

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## Feature Vector

Standard description for many visual processes:  
form a  from set of descriptions:

$$\vec{x} = (\text{compactness}, ci_1, ci_2, ci_3, ci_4, ci_5, ci_6)'$$

Multiple vectors if several structures or  
locations to describe

These vectors are then used in next processes,  
eg. recognition

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## Lecture Overview

1. Moments: an infinite family of shape descriptions
2. A way to make them  to rotation, translation, and scale

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