Image Capture and Problems	
	_
A reasonable capture	
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Image Capture: Shadow problems

False color to emphasize the shadow location. Often hard to separate from part.



Image Capture: Focus problems



Focus set to one distance. Nearby distances in focus (depth of focus). Further not so well focused. Compare 'identical' lines.

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Image Capture: Specularities/highlights



Saturated pixels set to red.

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Image Capture: Radial lens	
distortion	
nage Capture: Problems are Eccels are Eccels	
Note 'straight' lines at image edge. May make accurate measurements hard.	



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Image Capture: Overcoming Problems

- Shadows, specularities, non-uniform illumination: increase ambient lighting by using light diffusing panels or lots of point lights
- **Depth of Focus**: use smaller aperture and brighter light
- Motion Blur: use shorter capture time and brighter light
- **Saturation**: use smaller aperture, reduce gain and adjust gamma

- Lens Distortion: more expensive lenses, view from further away
- Aliasing: use incandescent lights

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Illumination control techniques

Main cause of problem: point light sources

Brightness = B / (surface distance from source)²



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Isolating flat parts

Isolate parts, then characterise later

Assume

- Dark part
- Light background
- Reasonably uniform illumination -> distinguishable parts

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Part and typical distribution





Spread: not quite uniform illumination + part color variations + sensor noise

Thresholding	
Thresholding: central technique	
<pre>for row = 1 : height for col = 1 : width if value(row,col) < ThreshHigh % inside high bnd % & value(row,col) > ThreshLow % optional low bnd output(row,col) = 1; else output(row,col) = 0;</pre>	
ena	

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Convolution

General purpose image (and signal) processing function

Computed by a weighted sum of image data and a fixed mask

Linear operator: $conv(a^*B,C) = a^*conv(B,C)$

Used in different processes: noise removal, smoothing, feature detection, differentiation, ...



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Histogram Smoothing for threshold selection

Histogram Smoothing (in findthresh.m) Convolve with a Gaussian smoothing window

filterlen = 50; % filter length
thefilter = gausswin(filterlen,sizeparam); % size=4
thefilter = thefilter/sum(thefilter); % unit norm
tmp2=conv(thefilter,thehist); % makes longer output
% select corresponding portion
offset = floor((filterlen+1)/2);
tmp1=tmp2(offset:len+offset-1);

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