Interpretation Tree Model Matching

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Interpretation Tree Model Matching

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Interpretation Tree matching

Goal: Correspondence between subset of Mmodel features $\{m_i\}$ and D data features $\{d_i\}$

Complete (exhaustive, depth-first) search - if a match exists, it will be found

Needs a 'wildcard' ('*') data feature to match model features with no corresponding data feature (occlusion, segmentation failure)

Can find multiple solutions

Result: $\{(m_i, d_{i_i})\}$ set of matched features

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Given:

Sets of model lines $\{m_i\}$ in a scene coordinate system Set of image lines $\{d_i\}$ in an image coordinate system Image to scene scale conversion factor σ (pixels to cm)

Do:

- 1. Match image and model lines $\{(m_i, d_i)\}$
- 2. Estimate transformation mapping model onto data: R, \vec{t}
- 3. Verify matching and pose estimate

Output: identity and position (R, \vec{t})

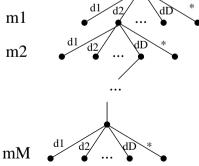
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Search Tree

Expand by model feature at each new level



Any given node in tree represents a set of matches $\{(m_i, d_{i_i})\}$

Reducing Search Complexity

Do we need to consider all paths in search tree? No: Suppose current match state has these pairs matched: $\{(m_i, d_{j_i})\}, i = 1..k$ Given a new pair $(m_{k+1}, d_{j_{k+1}})$

- 1. $unary_test(m_{k+1}, d_{j_{k+1}})$ terminates extending search path if new pair has incompatible properties
- 2. $binary_test(m_{k+1}, d_{j_{k+1}}, m_x, d_{j_x})$ for all x = 1..k terminates extending search path

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Computational Complexity

M model feature tree levels. D data features on each level plus 1 wildcard

Worst case: $(D+1)^M$ nodes in tree to visit

 p_u - probability that any random model feature and any random data feature pass unary_test

 p_b - probability that any 2 random model features and any 2 random data features pass binary_test

Then, if $p_bMD < 2$, then the average case complexity of ITREE search is $O(LD^2)$

Much smaller, but can still be substantial

if new pair has incompatible properties with each previous pairing on this tree branch (as all parts of the same object are compatible).

- 3. Early success limit L can stop search when have $\{(m_i, d_{j_i})\}, i = 1..L$ compatible pairs
- 4. Early failure limit L can stop search when can never get L pairs on this path. If have t non-wildcard matches on this path out of k pairings, then fail if t + (M k) < L

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IT algorithm matlab code

```
% interpretation tree - match model and data lines until
```

 $\ensuremath{\text{\%}}$ Limit are successfully paired or can never get Limit

% model - current model

% numM - number of lines in the model

% mlevel - last matched model feature

 $\mbox{\ensuremath{\mbox{\%}}}$ Limit - early termination threshold

% pairs(:,2) - paired model-data features

 $\mbox{\ensuremath{\mbox{\%}}}$ numpairs - number of paired features

function ok=itree(model,numM,mlevel,Limit,pairs,numpairs)

global Models numlines datalines

```
[theta,trans] = estimatepose(model,numpairs,pairs)
 for p = 1 : 4
    ok = verifymatch(theta(p),trans(p,:)',model,
              numpairs,pairs);
    if ok
                % successful verification
      return
    end
  end
                % failure to verify - continue search
  return
end
% never enough pairs
if numpairs + numM - mlevel < Limit</pre>
  ok=0;
  return
end
```

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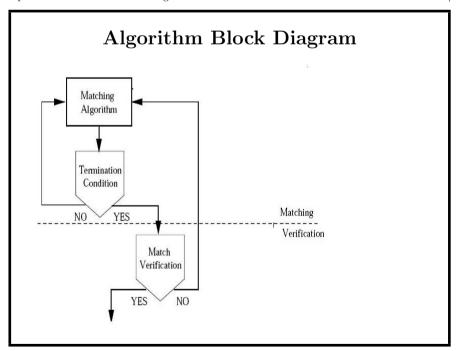
```
if passed
      \% passed all tests: add to matched pairs and recurse
      pairs(numpairs+1,1)=mlevel;
     pairs(numpairs+1,2)=d;
      ok=itree(model,numM,mlevel,Limit,pairs,numpairs+1);
      if ok
        return % successful verification
      end
    end
  end
end
% wildcard case - go to next model feature
ok = itree(model,numM,mlevel,Limit,pairs,numpairs);
```

```
% normal case - see if we can extend pair list
mlevel = mlevel+1;
for d = 1 : numlines
                       % try all data lines
 % do unary test
 if unarytest(model,mlevel,d)
   % do all binary tests
    passed=1;
   for p = 1 : numpairs
     if ~binarytest(model,mlevel,d,pairs(p,1),pairs(p,2))
         passed=0;
         break
     end
    end
```

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Line matching unary test

DATA LINE MODEL LINE



Pass test if $\sigma l_m (1 - \delta_u) \le l_d \le \sigma l_m (1 + \delta_u)$

Allows for calibration and segmentation errors Position independent property $(\delta_u = 0.3 \text{ typical})$

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Matching performance

Limit L = number of model lines - 1Tries all models Stops at first verified model instance for each model

Line matching binary tests

DATA LINES MODEL LINES

Pass test if $|\alpha - \beta| < \delta_b$

Allows for calibration and segmentation errors Position independent property $(\delta_b = 0.2 \text{ radians typical})$

Also: don't allow duplicate use of model or data lines

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Different Matched Models & Instances

Image	True Model	Tee	Thin L	Thick L
1	Tee	4	0	12
2	Tee	4	0	12
3	Tee	21	0	12
4	Tee	21	0	12
5	Thin L	0	15	2
6	Thin L	0	15	2
7	Thin L	0	15	2
8	Thin L	0	24	2
9	Thick L	0	2	3
10	Thick L	0	2	3
11	Thick L	0	2	3
12	Thick L	0	2	3

What Have We Learned?

Introduction to

- General Feature Matching Algorithm
- Efficient if good unary/binary tests
- Suitable for 50% (estimated) flat parts

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