1: Image Correlation

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Lecture 1: Image Correlation

Lecturer: MicMac/MPD/ER/JMM Scribes:

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1.1 Image Correlation

Correlation is a technique uses to evaluate "the similarity" or "the coherence" between 2 signals.

Lets begin with 2 image patch denoted f(x,y) and g(x,y), of same size N*M. To evaluate if 2 patchs are similar, we can compare pixel by pixel with the same position in 2 image patchs, and evaluate *euclidean distance* between each pair of pixel:

$$d_{ED} = \sqrt{\sum_{i=1}^{N} \sum_{j=1}^{M} (f(x_i, y_j) - g(x_i, y_j))^2}$$
(1.1)

This method give a result = 0 if 2 image patchs is exactly the same, but in another case, it is not so easy to evaluate the level of similarity, because the maximum value of d_{ED} depends on the number of pixel in patch and the maximum value of pixel also. And this method is so sensitive with small changes between 2 image patch like luminosity (brightness), a little deformation/translation, or noise.

So, lets turn to use the *correlation* as an indicator for similarity mesurement:

$$C_{fg} = \sum_{i=1}^{N} \sum_{j=1}^{M} (f(x_i, y_j) * g(x_i, y_j))$$
(1.2)

The higher the value C_{fg} , the better the similarity. Imagine that f is our template image patch, and we need to matching this template with target image G. So, we scan through all position of image G, extract an target image patch g at each position, and apply this equation to mesure similarity between 2 patchs f and g. All of this procedure means that we do a **cross-correlation**.

This equation is better with the minor different between 2 image patch in compare with d_{ED} . We can normalize the value of C_{fg} to keep the correlation coefficient in range [0, 1] and overcome the ilumination change problems. So, we have the *Normalize Cross-Correlation NCC*:

$$NCC_{fg} = \frac{\sum_{i=1}^{N} \sum_{j=1}^{M} (f(x_i, y_j) * g(x_i, y_j))}{\sqrt{\sum_{i=1}^{N} \sum_{j=1}^{M} f(x_i, y_j)^2 * \sum_{i=1}^{N} \sum_{j=1}^{M} g(x_i, y_j)^2}}$$
(1.3)

Explication about the different between Correlation & NCC: imagine that we evaluate correlation between image patch f(x, y) and $g_1(x, y)$ for the first time, and f(x, y) and $g_2(x, y)$ for the second time (Figure 1.2), with the constant factor intensity value change between g_1 and g_2 :

Our correlation and NCC coefficient between f and g_1 is:

$$C_{fg_1} = Aa + Bb + Cc + Dd + Ee + Ff + Gg + Hh + Ii$$

$$(1.4)$$

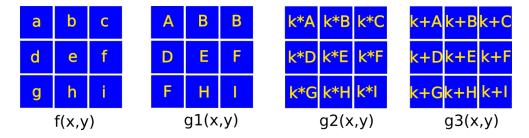


Figure 1.1: Correlation with a constant change pixel value patch

$$NCC_{fg_1} = \frac{Aa + Bb + Cc + Dd + Ee + Ff + Gg + Hh + Ii}{\sqrt{(A^2 + B^2 + C^2 + D^2 + E^2 + F^2 + G^2 + H^2 + I^2) * (a^2 + b^2 + c^2 + d^2 + e^2 + f^2 + g^2 + h^2 + i^2)}}$$
(1.5)

Our correlation and NCC coefficient between f and g_2 is:

$$C_{fg_2} = k * (Aa + Bb + Cc + Dd + Ee + Ff + Gg + Hh + Ii)$$
(1.6)

$$NCC_{fg_2} = \frac{k * (Aa + Bb + Cc + Dd + Ee + Ff + Gg + Hh + Ii)}{\sqrt{k^2(A^2 + B^2 + C^2 + D^2 + E^2 + F^2 + G^2 + H^2 + I^2) * (a^2 + b^2 + c^2 + d^2 + e^2 + f^2 + g^2 + h^2 + i^2)}}$$
(1.7)

So, the correlation coefficient is varies with constant illumination change if we compute with correlation (eq 1.2 gives $C_{fg_2} = k * C_{fg_1}$), but it rests equal if we take NCC value (eq 1.3 gives $NCC_{fg_2} = NCC_{fg_1}$).

But, if the changement of intensity value is a shifted constant value on the whole image patch, like from $g_1(x,y)$ to $g_3(x,y)$ (Figure 1.1), both correlation and NCC value is varies also. To minimize the effect of this problem on correlation score, we normalize each pixel intensity and image patch standard deviation value with the image patch mean intensity value before compute correlation:

$$ZNCC_{fg} = \frac{\sum_{i=1}^{N} \sum_{j=1}^{M} (f(x_i, y_j) - \bar{f}) * (g(x_i, y_j) - \bar{g})}{\sqrt{\sum_{i=1}^{N} \sum_{j=1}^{M} (f(x_i, y_j) - \bar{f})^2 * \sum_{i=1}^{N} \sum_{j=1}^{M} (g(x_i, y_j) - \bar{g})^2}}$$
(1.8)

Equation 1.8 called **Zero Mean Normalized Cross Correlation**. It gives value varies in [0,1] with the higher is the better correlate.

1.2 Implementation in MicMac

We try to implemented in MicMac a little program that find a best matched position of a little template (image patch) in a large target image.

Our very simple test describe below:

- Take an image (target)
- Cut an image patch, use it as a matching template.
- We try to re-find the position of image patch in taget.

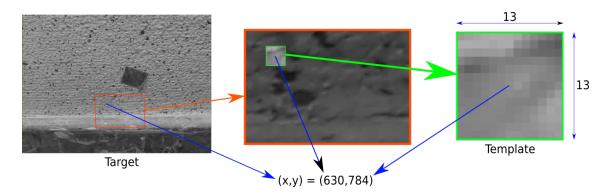


Figure 1.2: Template taken from image (target) and ground truth position

The program is implemented in "micmac/src/TpMMPD/Ex_Match", with command "mm3d TestLib LSQ-Match".

1.2.1 Using class RMat_Inertie

The correlation compute part is written in cLSQTemplate.cpp with 2 sub function:

- dblTst_Correl1Win: compute correlation score between 2 image patch.
- dblTst_Correl: search for best matched position between template image patch and target image by compute correlation score for template image patch with all position in target image, by calling dblTst_Correl1Win.

```
double dblTst_Correl1Win
     tIm2DM & Im1,
                                     // image 1
     const Pt2dr & aP1,
                                        center coordinate of image patch 1
     tIm2DM & Im2,
                                             // image 2
     const Pt2dr & aP2,
                                        center coordinate of image patch 2
                                     // size of correlation windows
     const Pt2dr aSzW,
                                     // pixel sampling step
     const int aStepPxl,
     {\color{red} \mathbf{const}} \hspace{0.2cm} \mathbf{cInterpolateurIm2D} {\color{red} < } \mathbf{double} {\color{blue} > \&} \hspace{0.2cm} \mathbf{aInterPol}
                                                                 //an interpolater
    /* Compute point most up & most down for two images patch
     * check if we can take the patch with coordinate and windows size given
    Pt2dr aPtSupIm1 = aP1 + aSzW;
    Pt2dr aPtInfIm1 = aP1 - aSzW;
    Pt2dr aPtSupIm2 = aP2 + aSzW;
    Pt2dr aPtInfIm2 = aP2 - aSzW;
       if one of these point lie outside the image => quit program
    if (!(
              (InsideREAL(Im1, aPtSupIm1) && InsideREAL(Im1, aPtInfIm1))
              (InsideREAL(Im2, aPtSupIm2) && InsideREAL(Im2, aPtInfIm2))
        ) )
              return (TT_DefCorrel);
     double aDefInter = -1.0; // default value for interpolater
```

```
cResulRechCorrel dblTst_Correl
 tIm2DM & Im1,
                       // image 1
 const Pt2dr & aP1,
                       // center coordinate of image patch 1
                       // image 2
 tIm2DM & Im2,
                       // center coordinate of search zone on image 2
 const Pt2dr & aP2,
 const Pt2dr aSzW,
                      // size of correlation windows
 const Pt2dr aSzRech,// size of search zone
                     // score correlation image (result)
 tIm2DM & ImScore,
 const cInterpolateurIm2D<double> & aInterPol, // interpolator for pixel sampling
 bool & OK
                    // if everythings is ok
   /* Compute point most up & most down for two images patch
    * check if we can take the patch with coordinate and windows size given
   Pt2dr aPtSupIm1 = aP1 + aSzW;
   Pt2dr aPtInfIm1 = aP1 - aSzW;
    // taken in account some reserve for interpolater
   int aSzInterpol = aInterPol.SzKernel();
   Pt2dr aPtSupIm2 = aP2 + aSzRech + Pt2dr(aSzInterpol, aSzInterpol);
   Pt2dr aPtInfIm2 = aP2 - aSzRech - Pt2dr(aSzInterpol, aSzInterpol);
   cout <<a PtSupIm1 <<a PtInfIm1 << "_" <<a PtInfIm2 << end1;
    if (!(
            (InsideREAL(Im1, aPtSupIm1) && InsideREAL(Im1, aPtInfIm1))
           (InsideREAL(Im2, aPtSupIm2) && InsideREAL(Im2, aPtInfIm2))
      ) )
   {
           OK=false:
           return (cResulRechCorrel (Pt2dr(-1,-1),TT_DefCorrel));
   }
   double aScoreMax = -1e30;
   \begin{array}{ll} Pt2dr & aDecMax\,; \\ Pt2dr & aP\,; \end{array}
    // sweep the correlation windows over image 2
   for (aP.x = -aSzRech.x ; aP.x < aSzRech.x ; aP.x = aP.x + aStep)
        for (aP.y=-aSzRech.y; aP.y<= aSzRech.y; aP.y = aP.y + aStep)
             // compute correlation score
            double a2Sol = dblTst_Correl1Win(Im1, aP1, Im2, aP2+aP, aSzW, aStepPxl, aInterPol);
```

Let see how the class RMat_Inertie compute correlation score. This class is created from class Mat_Inertie with value type double. Class is defined and implemented in "micmac/include/general/geom_vecteur.h".

```
* class variable member and constructor
Mat_Inertie
  ElTyName Type::TypeScal S,
  ElTyName Type::TypeEff S1,
  ElTyName Type::TypeEff S2,
  ElTyName Type::TypeScal S11,
  ElTyName Type::TypeScal S12,
ElTyName Type::TypeScal S22
 )
  _{\rm -S}
                 // _s: number of pixel pair added
           (S1), // _{\rm s}1: sum of all (v1), with v1 is intensity value of pixel image 1 (S2), // _{\rm s}2: sum of all (v2), with v2 is intensity value of pixel image 2 (S11), // _{\rm s}11: sum of all (v1)*(v1) => standard deviation of img 1 (S12), // _{\rm s}11: sum of all (v1)*(v2)
  _{\mathtt{s}}1
  _{\rm s}2
  _{s}11
  _{\mathtt{s}}12
  _{\rm s}22
           (S22) // _s22: sum of all (v2)*(v2) => standard deviation of img 2
  {
/* Pre-compute needed value for each pixel pair added */
void add_pt_en_place(ElTyName Type::TypeEff v1,ElTyName Type::TypeEff v2)
                   += 1; // increase counter of number of pixel pair added
               _{s1} += v1;
               _{s2} += v2;
               _s11 += scal(v1,v1); // scal(v1,v1)=(v1)*(v1)
               -s12 += scal(v1, v2);
               _{s22} += scal(v2, v2);
        }
/* Normalize pixel value */
Mat_Inertie < ElTyName Type:: TypeReel > normalize() const
        // check if there are pixel to normalize
                ELISE_ASSERT
                (
                       _s != 0,
                      "som_pds_=_0_in_Mat_Inertie::normalize"
                );
```

1.2.2 Using class cCorrelImage

This is another class use to compute correlation score also (Jean-Michael Muller). In this class, the implementation idea is 1 object of this class is 1 image patch. Have a look at class declaration:

```
/* Idea General: 1 object of this class is 1 image patch */
class cCorrelImage
  public :
    cCorrelImage();
    /* Im2D and TIm2D is 2 image type usually use in MicMac.
       The different ?Hmmmm....TIm2D let pixel access faster in some case...
     * But TIm2D can declare from an Im2D object.
     * So, we will usually see these 2 type comes together
    Im2D < U_INT1, INT4 > * getIm() \{ return \& mIm; \}
   \label{eq:tim2D} $$TIm2D<U_INT1,INT4>* getImT(){return &mTIm;}$$// compute correlation coeff between this image patch & aIm2 image patch}$
    double CrossCorrelation(const cCorrelImage & aIm2);
    /* compute covariance b/w this image patch & aIm2 image patch:
    * cov = sigma(f*g)/(sqrt(1+2n))
    double Covariance(const cCorrelImage & aIm2);
    // mSzW => half size of correlation windows
    int getSzW();
     / mSz => size of correlation windows
    Pt2di getmSz();
    /* extract image patch with size = mSz at position (aCenterX, aCenterY)
     * from a large image anIm
    void \ getFromIm(Im2D\!<\!U\_INT1,INT4\!>\ *\ anIm\,, \\ double \ aCenterX\,, \\ double \ aCenterY\,);
     / copy whole image anIm as image patch
    void getWholeIm(Im2D<U_INT1,INT4> * anIm);
    // Set size for correlation windows
```

```
static void setSzW(int aSzW);
static int mSzW;

protected:
    // pre-compute value needed for correlation => compute mImS1 & mImS2
    void prepare();
    Pt2di mSz;
    // this image patch in 2 types Im2D & TIm2D
    TIm2D<U_INT1,INT4> mTIm;
    Im2D<U_INT1,INT4> mIm;
    // each pixel in mIm with their value v1 becomes v1/(1+2n)
    TIm2D<REAL4,REAL8> mTImS1;
    Im2D<REAL4,REAL8> mImS1;
    // each pixel in mIm with their value v1 becomes (v1)^2/(1+2n)
    TIm2D<REAL4,REAL8> mTImS2;
    Im2D<REAL4,REAL8> mTImS2;
    Im2D<REAL4,REAL8> mImS2;
};
```

To use this class, a little code is written in cLSQTemplate.cpp, in "main" function, part "Corrrelation by cCorrelImage". Let see and explication how to use it:

```
typedef Im2D<unsigned char, int> tIm2DcCorrel;
typedef TIm2D<unsigned char, int> tTIm2DcCorrel;
  cResulRechCorrel just for contain correlation result (coordinate & score)
cResulRechCorrel aResCorrel(Pt2dr(-1,-1),TT_DefCorrel);
// create correlation result image
tIm2DM aImgScoreCorrel(aImgTarget->Im2D().sz().x, aImgTarget->Im2D().sz().y);
// Load Image to type Im2D<unsigned int, int> \Rightarrow class cCorrelImage require this image type
Pt2di aSzImTarget = aImgTarget->Im2D().sz();
Pt2di aSzImTmp = aImgTmplt->Im2D().sz();
tIm2DcCorrel aCImTmpl(aSzImTmp.x, aSzImTmp.y);
tIm2DcCorrel aCImTarget(aSzImTarget.x, aSzImTarget.y);
ELISE_COPY(aCImTmpl.all_pts(), aImgTmplt->Tif().in(), aCImTmpl.out());
ELISE_COPY(aCImTarget.all_pts(), aImgTarget->Tif().in(), aCImTarget.out());
\ast At this moment, we have loaded a template image patch aCImTmpl
 * and a full target image aCImTarget, with the type used for class cCorrelImage.
 * Now, we will create a template image patch as object of class cCorrelImage
 * and scan through target image aCImTarget, creat a target image patch as object
 * of class cCorrelImage to compute correlation score at each position.
 */
// Before using cCorrelImage class, we need to set image patch size (half size)
cCorrelImage::setSzW(aSzW.x);
 / Now, when we create a new image patch object, it will initialize with size = aSzW*2+1
cCorrelImage ImPatchTmpl;
// Take all template image as template image patch
ImPatchTmpl.getWholeIm(&aCImTmpl);
// Create a target image patch, size aSzW*2+1, for now it's empty
cCorrelImage ImPatchTarget;
bool corOK = false;
Pt2dr aPt(0,0);
OK = true;
         // Now, scan through target image, each pixel position:
         for (aPt.x=0; aPt.x<aSzImTarget.x; aPt.x = aPt.x + (int)aParam.mStepCorrel)
             for (aPt.y=0; aPt.y<aSzImTarget.y; aPt.y = aPt.y + (int)aParam.mStepCorrel)
             {
                 Pt2dr aPtInf = aPt-Pt2dr(ImPatchTmpl.getmSz());
                    if image patch is inside image
                         (aPtInf.x >= 0 \&\& aPtInf.y >= 0)
                      && (aPtInf.x < aSzImTarget.x && aPtInf.y < aSzImTarget.y) )
```

```
// Load image patch at position aPt from target image
                       ImPatchTarget.getFromIm(\&aCImTarget,\ aPt.x,\ aPt.y);
                        // compute correlation score b/w 2 image patch
                       double aScore = ImPatchTmpl.CrossCorrelation(ImPatchTarget);
                       corOK = true;
                       if (aScore > aResCorrel.mCorrel)
                             update best match position
                            aResCorrel.mCorrel = aScore;
                            aResCorrel.mPt = Pt2dr(aPt);
                        // write score to correlation result image
                       aImgScoreCorrel.SetR_SVP(Pt2di(aPt), aScore);
                  else
                       corOK = false;
                       aImgScoreCorrel.SetR\_SVP(Pt2di(aPt), -1.0);
         }
if (OK)
  cout << "Correl .: _" << aResCorrel . mCorrel << " _ _ Pt : _" << aResCorrel . mPt << endl ;
    cout << " Correl_false "<< endl;</pre>
```

Execute command that implemente this example:

```
mm3d TestLib LSQMatch Temp_13_Test.png Target.png StepCor=1 NbIter=0 Meth=1
```

Params: "StepCor=1" is the movement step of correlation windows. "Meth=1" to tell the program compute correlation by class cCorrelImage. We obtain result:

```
Correl: 0.99999 - Pt: [630,784]
```

which is exact match (view ground truth and dataset in Figure 1.2)

1.2.3 Using class Optim2DParam

1.2.4 **CENCUS**

Function CencusBasic, CencusBasicCenter (regarder Param XML_MICMAC.xml, search Cencus),

MutualInformation (pas dans micmac): Similarity=(Entropy(g) *entropy(f))/(Entropy(g*f))