Efficient neighbourhood computing for discrete rigid transformation graph search

Yukiko Kenmochi Phuc Ngo Hugues Talbot Nicolas Passat









September 10, 2014

4 D > 4 D > 4 E > 4 E > E *) Q (

1/25

Background

■ Problem of rigid image registration is usually formulated in the **continuous** domain, in the context of optimisation.



Background

- Problem of rigid image registration is usually formulated in the continuous domain, in the context of optimisation.
- This often requires **re-digitizing** results after transformation, and sometimes causes unwanted artifacts.

Background

- Problem of rigid image registration is usually formulated in the continuous domain, in the context of optimisation.
- This often requires **re-digitizing** results after transformation, and sometimes causes unwanted artifacts.

Motivation and clues

Is it possible to avoid this re-digitization?

Background

- Problem of rigid image registration is usually formulated in the continuous domain, in the context of optimisation.
- This often requires **re-digitizing** results after transformation, and sometimes causes unwanted artifacts.

Motivation and clues

Is it possible to avoid this re-digitization?

Yes, a **fully discrete approach** allows to transform images pixel by pixel.

Background

- Problem of rigid image registration is usually formulated in the continuous domain, in the context of optimisation.
- This often requires re-digitizing results after transformation, and sometimes causes unwanted artifacts.

Motivation and clues

- Is it possible to avoid this re-digitization?
 - Yes, a **fully discrete approach** allows to transform images pixel by pixel.
- How to explore efficiently such a discrete parameter space?



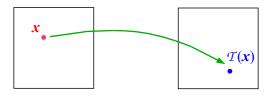
Rigid transformation on \mathbb{R}^2

Definition (Rigid transformation $\mathcal{T}: \mathbb{R}^2 \to \mathbb{R}^2$)

A rigid transformation is a bijection defined for any $\mathbf{x} = (x, y) \in \mathbb{R}^2$ as

$$\mathcal{T}(\mathbf{x}) = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

with $a_1, a_2 \in \mathbb{R}$ and $\theta \in [0, 2\pi[$.



Rigid transformation on \mathbb{Z}^2

Definition (Digital rigid transformation $T: \mathbb{Z}^2 \to \mathbb{Z}^2$)

A digital rigid transformation on \mathbb{Z}^2 is defined for any $oldsymbol{p}=(p,q)\in\mathbb{Z}^2$ as

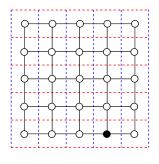
$$T(\mathbf{p}) = D \circ T(\mathbf{p}) = \begin{pmatrix} [p\cos\theta - q\sin\theta + a_1] \\ [p\sin\theta + q\cos\theta + a_2] \end{pmatrix}$$

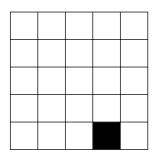
where $D: \mathbb{R}^2 \to \mathbb{Z}^2$ is a digitization (a rounding function).



Discontinuities of rigid transformations in \mathbb{Z}^2

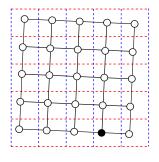
$$T(\mathbf{p}) = D \circ T(\mathbf{p}) = \begin{pmatrix} [p\cos\theta - q\sin\theta + a_1] \\ [p\sin\theta + q\cos\theta + a_2] \end{pmatrix}$$

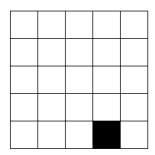




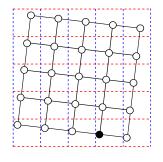
Discontinuities of rigid transformations in \mathbb{Z}^2

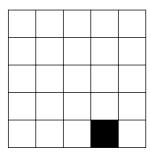
$$T(\mathbf{p}) = D \circ T(\mathbf{p}) = \begin{pmatrix} [p\cos\theta - q\sin\theta + a_1] \\ [p\sin\theta + q\cos\theta + a_2] \end{pmatrix}$$





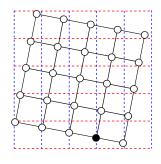
$$T(\mathbf{p}) = D \circ T(\mathbf{p}) = \begin{pmatrix} [p\cos\theta - q\sin\theta + a_1] \\ [p\sin\theta + q\cos\theta + a_2] \end{pmatrix}$$

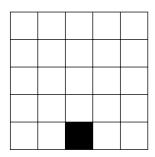




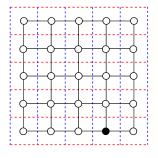
Discontinuities of rigid transformations in \mathbb{Z}^2

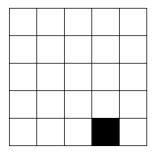
$$T(\mathbf{p}) = D \circ T(\mathbf{p}) = \begin{pmatrix} [p\cos\theta - q\sin\theta + a_1] \\ [p\sin\theta + q\cos\theta + a_2] \end{pmatrix}$$



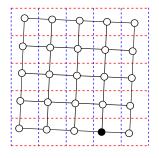


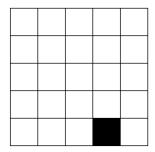
Definition



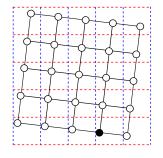


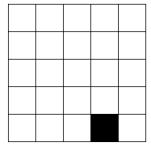
Definition



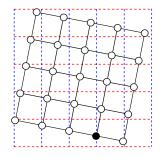


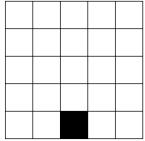
Definition





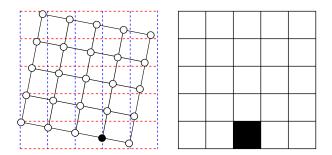
Definition





Definition

A discrete rigid transformation (DRT) is the set of all the rigid transformations that generate a same image.

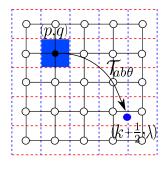


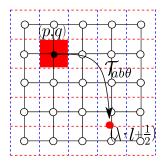
The parameter space (a_1, a_2, θ) is partitioned by the disjoint set of DRTs.

Critical rigid transformations

Definition

A critical rigid transformation moves at least one point of \mathbb{Z}^2 to a point on the vertical or horizontal half-grid.





Tipping surfaces

Definition

The **tipping surfaces** are the surfaces associated to critical transformations in the parameter space (a_1, a_2, θ) :

$$a_1 = k + \frac{1}{2} + q \sin \theta - p \cos \theta,$$
 (vertical)

$$a_2 = I + \frac{1}{2} - p \sin \theta - q \cos \theta,$$
 (horizontal)

for $p, q, k, l \in \mathbb{Z}$.

Tipping surfaces

Definition

The **tipping surfaces** are the surfaces associated to critical transformations in the parameter space (a_1, a_2, θ) :

$$a_1 = k + \frac{1}{2} + q \sin \theta - p \cos \theta,$$
 (vertical)

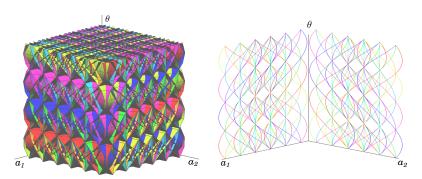
$$a_2 = l + \frac{1}{2} - p \sin \theta - q \cos \theta,$$
 (horizontal)

for $p, q, k, l \in \mathbb{Z}$.

Each tipping surface

- is indexed by a **triplet of integers** (p, q, k) (resp. (p, q, l)),
- indicates that the pixel (p, q) in a transformed image **changes its** value from the one at (k, *) (resp. (*, I)) in an original image to the one at (k + 1, *) (resp. (*, I + 1)).

Example of tipping surfaces



Vertical surfaces Φ_{pqk} and horizontal ones Ψ_{pql} for $p, q \in [0, 2]$ and $k, l \in [0, 3]$.

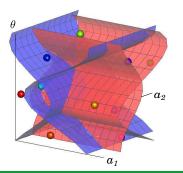
DGCI 2014 9/25

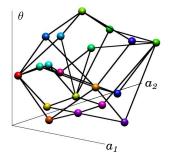
Graph of discrete rigid transformations

Definition

A graph of discrete rigid transformations (DRT graph) is a graph G = (V, E) such that :

- lacksquare each vertex $v \in V$ corresponds to a DRT,
- lacktriangle each edge $e \in E$ connects two DRTs sharing a tipping surface.





■ *) Q (*

Advantages

 Discrete rigid transformations are computed in a fully discrete process.

Advantages

- Discrete rigid transformations are computed in a fully discrete process.
- Their combinatorial structure is represented by a DRT graph G whose complexity is $O(N^9)$ for images of size $N \times N$.

Advantages

- Discrete rigid transformations are computed in a fully discrete process.
- Their combinatorial structure is represented by a DRT graph G whose complexity is $O(N^9)$ for images of size $N \times N$.
- *G* models all the discrete rigid transformations with their **topological information** such that :
 - a vertex corresponds to a different transformed image,
 - an edge corresponds to one pixel change, i.e. a tipping surface, (each edge possesses such a pixel transition information).

Advantages

- Discrete rigid transformations are computed in a fully discrete process.
- Their combinatorial structure is represented by a DRT graph G whose complexity is $O(N^9)$ for images of size $N \times N$.
- G models all the discrete rigid transformations with their topological information such that :
 - a vertex corresponds to a different transformed image,
 - an edge corresponds to one pixel change, *i.e.* a tipping surface, (each edge possesses such a pixel transition information).
- It enables to generate exhaustively and incrementally all the transformed images in linear time.

Problem formulation

Given two images A and B of size $N \times N$, image registration consists of finding a DRT such that

$$v^* = \arg\min_{v \in V} d(A, T_v(B))$$

where T_v is the digital rigid transformation of a DRT v, and d is a given distance between two images.

Problem formulation

Given two images A and B of size $N \times N$, image registration consists of finding a DRT such that

$$v^* = \arg\min_{v \in V} d(A, T_v(B))$$

where T_v is the digital rigid transformation of a DRT v, and d is a given distance between two images.

We have a choice for d; here we use **signed distance**. (Boykov *et al.*, 2006)

Problem formulation

Given two images A and B of size $N \times N$, image registration consists of finding a DRT such that

$$v^* = \arg\min_{v \in V} d(A, T_v(B))$$

where T_v is the digital rigid transformation of a DRT v, and d is a given distance between two images.

We have a choice for d; here we use **signed distance**. (Boykov et al., 2006)

Disadvantage

Exhaustive search on DRT graph costs $O(N^9)$ in complexity.

Problem formulation

Given two images A and B of size $N \times N$, image registration consists of finding a DRT such that

$$v^* = \arg\min_{v \in V} d(A, T_v(B))$$

where T_{ν} is the digital rigid transformation of a DRT ν , and d is a given distance between two images.

We have a choice for d; here we use **signed distance**. (Boykov et al., 2006)

Disadvantage

Exhaustive search on DRT graph costs $O(N^9)$ in complexity.

Advantage

A local search on DRT graph can determine a local optimum.

Local search on discrete rigid transformation graph

Local search

- **Input** : An initial DRT $v_0 \in V$.
- **Output** : A local optimum $\hat{v} \in V$.
- **Approach** : Gradient descent : find a better solution in neighbours.

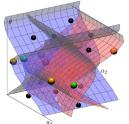
Local search on discrete rigid transformation graph

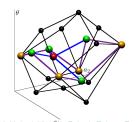
Local search

- **Input** : An initial DRT $v_0 \in V$.
- **Output** : A local optimum $\hat{v} \in V$.
- **Approach** : Gradient descent : find a better solution in neighbours.

DRT graph provides

- **neighbourhood structure** N(v)
- efficient computation of d





Local search on discrete rigid transformation graph

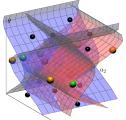
Local search

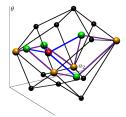
- **Input** : An initial DRT $v_0 \in V$.
- **Output**: A local optimum $\hat{v} \in V$.
- **Approach** : Gradient descent : find a better solution in neighbours.

DRT graph provides

■ neighbourhood structure $N^k(v) = N^{k-1}(v) \cup \bigcup_{u \in N^{k-1}(v)} N(u)$

efficient computation of d





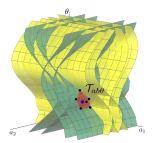
k-neighbourhood construction in a DRT graph

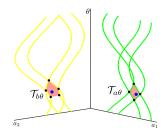
Around a DRT v, we can consider $4N^2$ inequalities :

$$p'_i - \frac{1}{2} < p_i \cos \theta - q_i \sin \theta + a_1 < p'_i + \frac{1}{2}$$

 $q'_i - \frac{1}{2} < p_i \sin \theta + q_i \cos \theta + a_2 < q'_i + \frac{1}{2}$

due to the pixel correspondence between (p_i, q_i) in the transformed image and (p'_i, q'_i) in the original image.





k-neighbourhood construction in a DRT graph

Around a DRT v, we can consider $4N^2$ inequalities :

$$p'_{i} - \frac{1}{2} < p_{i} \cos \theta - q_{i} \sin \theta + a_{1} < p'_{i} + \frac{1}{2}$$

$$q'_{i} - \frac{1}{2} < p_{i} \sin \theta + q_{i} \cos \theta + a_{2} < q'_{i} + \frac{1}{2}$$

due to the pixel correspondence between (p_i, q_i) in the transformed image and (p'_i, q'_i) in the original image.

k-neighbourhood construction problem

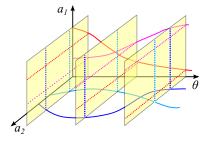
■ **Input** : 4N² tipping surfaces around v

• Output :
$$N^k(v) = N^{k-1}(v) \cup \bigcup_{u \in N^{k-1}(v)} N(u)$$

Previous time complexity $O(m^k N^2)$ is improved to $O(mkN^2)$ where m is vertex degree.

Sweeping plane algorithm for DRT sub-graph construction

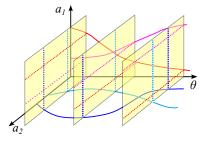
DRT sub-graph construction by sweeping plane algorithm, with 2 vertical and 2 horizontal tipping surfaces.



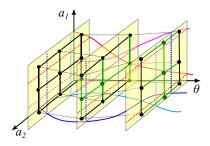
(a) 3×3 cells generated by tipping surfaces in each plane

Sweeping plane algorithm for DRT sub-graph construction

DRT sub-graph construction by sweeping plane algorithm, with 2 vertical and 2 horizontal tipping surfaces.



(a) 3×3 cells generated by tipping surfaces in each plane



(b) Associated DRT graph in each plane

15/25

- **Input** : a DRT v and its initial position θ_v .
- *Output* : The k-neighbours of v
- 1 Set $\theta_{prev} = \theta_v$.

- **Input** : a DRT v and its initial position θ_v .
- *Output* : The k-neighbours of v
- 1 Set $\theta_{prev} = \theta_v$.
- **2** Find 4(k+1) closest tipping surfaces around v.

- Input : a DRT v and its initial position θ_v .
- *Output* : The k-neighbours of v
- **2** Find 4(k+1) closest tipping surfaces around v.
- 3 Sort and put them in an ordered set S.

- Input : a DRT v and its initial position θ_v .
- *Output* : The k-neighbours of v
- **2** Find 4(k+1) closest tipping surfaces around v.
- 3 Sort and put them in an ordered set S.
- **4** Find the interval $[\theta_{prev}, \theta_{next}]$ where S is not changed.

- Input : a DRT v and its initial position θ_v .
- *Output* : The k-neighbours of v
- **2** Find 4(k+1) closest tipping surfaces around v.
- 3 Sort and put them in an ordered set S.
- **4** Find the interval $[\theta_{prev}, \theta_{next}]$ where S is not changed.
- **5** Sweep the plane from θ_{prev} to θ_{next} .

- **Input** : a DRT v and its initial position θ_v .
- Output : The k-neighbours of v
- **2** Find 4(k+1) closest tipping surfaces around v.
- **3** Sort and put them in an ordered set *S*.
- **4** Find the interval $[\theta_{prev}, \theta_{next}]$ where S is not changed.
- **5** Sweep the plane from θ_{prev} to θ_{next} .
- 6 If the neighbourhood depth of each generated vertex is not more than k, update S and go back to Step 4.

- Input : a DRT v and its initial position θ_v .
- Output : The k-neighbours of v
- **2** Find 4(k+1) closest tipping surfaces around v. $O(N^2)$
- 3 Sort and put them in an ordered set S. $O(k \log k)$
- **4** Find the interval $[\theta_{prev}, \theta_{next}]$ where S is not changed. $O(N^2)$
- **S** Sweep the plane from θ_{prev} to θ_{next} .
- If the neighbourhood depth of each generated vertex is not more than k, update S and go back to Step 4. O(1)

Algorithm (k-neighbourhood construction (in left-hand side of θ -axis))

- **Input** : a DRT v and its initial position θ_v .
- *Output* : The k-neighbours of v
- **2** Find 4(k+1) closest tipping surfaces around v. $O(N^2)$
- Sort and put them in an ordered set S. $O(k \log k)$
- 4 Find the interval $[\theta_{prev}, \theta_{next}]$ where S is not changed. $O(N^2)$
- **5** Sweep the plane from θ_{prev} to θ_{next} . $O(k^2)$
- If the neighbourhood depth of each generated vertex is not more than k, update S and go back to Step 4. O(1)

Estimated number of iterations (Steps 4-6) : m(2k + 1)

Algorithm (k-neighbourhood construction (in left-hand side of θ -axis))

- **Input** : a DRT v and its initial position θ_v .
- Output : The k-neighbours of v
- **2** Find 4(k+1) closest tipping surfaces around v. $O(N^2)$
- Sort and put them in an ordered set S. $O(k \log k)$
- 4 Find the interval $[\theta_{prev}, \theta_{next}]$ where S is not changed. $O(N^2)$
- **5** Sweep the plane from θ_{prev} to θ_{next} . $O(k^2)$
- If the neighbourhood depth of each generated vertex is not more than k, update S and go back to Step 4. O(1)

Estimated number of iterations (Steps 4-6) : m(2k + 1)**Total complexity** : $O(mkN^2)$

Let v, e and f be the numbers of vertices, edges and cells of the 2D DRT graph, which is the projected DRT graph onto the (a_i, θ) plane, i = 1, 2.

Let v, e and f be the numbers of vertices, edges and cells of the 2D DRT graph, which is the projected DRT graph onto the (a_i, θ) plane, i = 1, 2.

From

■ the Euler formula : v - e + f = 2,

Let v, e and f be the numbers of vertices, edges and cells of the 2D DRT graph, which is the projected DRT graph onto the (a_i, θ) plane, i = 1, 2.

From

- the Euler formula : v e + f = 2,
- the DRT graph property : $4f \le 2e$,

Let v, e and f be the numbers of vertices, edges and cells of the 2D DRT graph, which is the projected DRT graph onto the (a_i, θ) plane, i = 1, 2.

From

- the Euler formula : v e + f = 2,
- the DRT graph property : $4f \le 2e$,

we have the average degree

$$\frac{2e}{v} \le 4 - \frac{8}{v} < 4$$

as $v \gg 8$ in the 2D DRT graph, and thus

Let v, e and f be the numbers of vertices, edges and cells of the 2D DRT graph, which is the projected DRT graph onto the (a_i, θ) plane, i = 1, 2.

From

- the Euler formula : v e + f = 2,
- the DRT graph property : $4f \le 2e$,

we have the average degree

$$\frac{2e}{v} \le 4 - \frac{8}{v} < 4$$

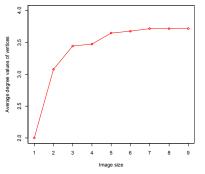
as $v \gg 8$ in the 2D DRT graph, and thus

Property

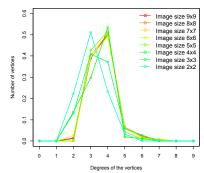
The average vertex degree of DRT graphs is lower than 8.

< ロ > ∢ 昼 > ∢ 差 > ∢ 差 > 差 | りへ(^)

Average vertex degrees of 2D DRT graphs in practice

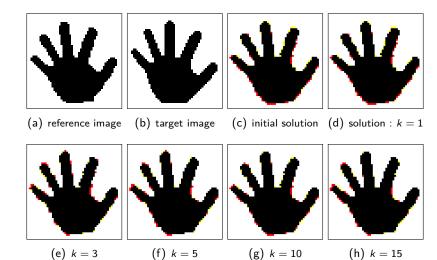


(a) Average vertex degree in a 2D DRT graph



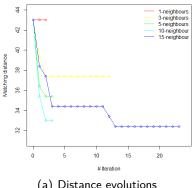
(b) Normalized vertex degree distribution in a 2D DRT graph

Experiment on binary images

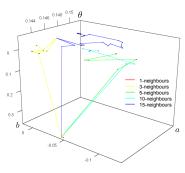


19/25

Experiment on binary images : distance evolutions

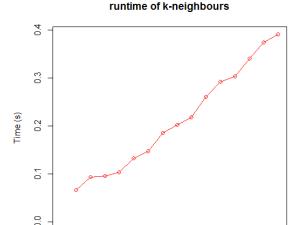


(a) Distance evolutions



(b) Transformation sequences

Experiment on binary images : runtime complexity



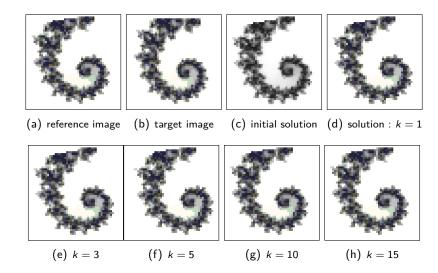
10

k-neighbour

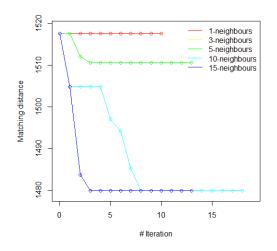
15

Experiment on binary images : image sequences for k = 15

Experiment on gray images



Experiment on gray images : distance evolutions



Conclusion

■ A purely discrete framework for 2D image registration under rigid transformation was proposed, based on DRT graph.



Conclusion

- A purely discrete framework for 2D image registration under rigid transformation was proposed, based on DRT graph.
- A gradient descent procedure constructs the exact k-neighbourhood at each step with a $O(mkN^2)$ time complexity.

Conclusion

- A purely discrete framework for 2D image registration under rigid transformation was proposed, based on DRT graph.
- A gradient descent procedure constructs the exact k-neighbourhood at each step with a $O(mkN^2)$ time complexity.
- The average vertex degree *m* is less than 8.



Conclusion

- A purely discrete framework for 2D image registration under rigid transformation was proposed, based on DRT graph.
- A gradient descent procedure constructs the exact k-neighbourhood at each step with a $O(mkN^2)$ time complexity.
- The average vertex degree *m* is less than 8.
- An efficient algorithm for computing the k-neighbours for the local search was proposed with linear time complexity.

Conclusion

- A purely discrete framework for 2D image registration under rigid transformation was proposed, based on DRT graph.
- A gradient descent procedure constructs the exact k-neighbourhood at each step with a $O(mkN^2)$ time complexity.
- The average vertex degree *m* is less than 8.
- An efficient algorithm for computing the k-neighbours for the local search was proposed with linear time complexity.

Perspectives

• Combine our proposed method with other combinatorial approaches.

DGCI 2014 25/25

Conclusion

- A purely discrete framework for 2D image registration under rigid transformation was proposed, based on DRT graph.
- A gradient descent procedure constructs the exact k-neighbourhood at each step with a $O(mkN^2)$ time complexity.
- The average vertex degree *m* is less than 8.
- An efficient algorithm for computing the k-neighbours for the local search was proposed with linear time complexity.

Perspectives

- Combine our proposed method with other combinatorial approaches.
- Extend to higher dimensions, and gray-level or labelled images.

DGCI 2014 25/25