An Acceleration of a Graph Cut Segmentation With FPGA

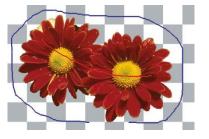
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What Is Graph Cut Segmentation?

- Graph cut is one of the segmentation methods based on energy minimization, and graph cut based segmentation is widely used.
- The following images are examples of segmentation [*].
- The seed pixels (target objects or background) are given by the user, and then only the target objects are extracted.



Input image with seeds



Output image

[*] Tomoyuki Nagahashi, Hironobu Fujiyoshi, and Takeo Kanade, "Image Segmentation Using Iterated Graph Cuts Based on Multi-scale Smoothing," in *ACCV 2007, Part II, LNCS 4844*, pp. 806--816, 2007.

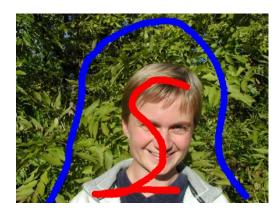
Background

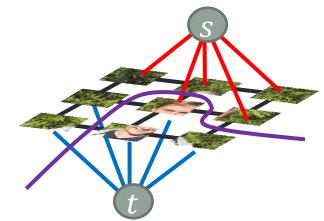
- For calculating the graph cut, max-flow algorithm is widely used, but it requires long computation time.
- We need an acceleration by FPGA or GPU for real-time processing of the max-flow algorithm.
- The performance of a GPU (GeForce GTX280) system [7] is 25 graph cuts per second on 640 x 480 pixel images, which is about 5 times faster than CPU.

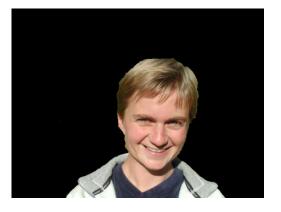
[7] V. Vineet and P. J. Narayanan, "Cudacuts: Fast graph cuts on the gpu," in *CVPR Workshop on Visual Computer Vision on GPUs*, 2008.

Segmentation Procedure

- Seed pixels (on objects or background) are specified by the user.
- A weighted directed graph among the pixels in the image is generated based on the seed pixels.
- A min-cut of the weighted directed graph is calculated using max-flow algorithm.





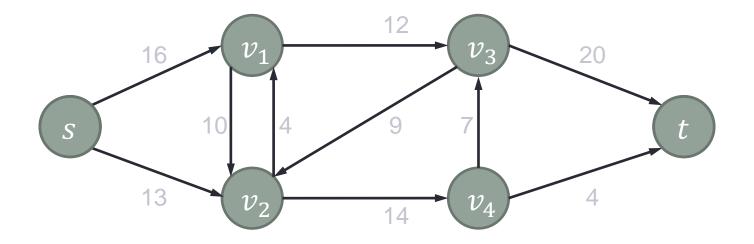


A Graph

• G = (V, E) : a weighted directed graph

V is a set of vertices (pixels), and it includes two special nodes, *s* and *t*.

E is a set of edges between two vertices, and each edge has a non-negative capacity c(u, v).

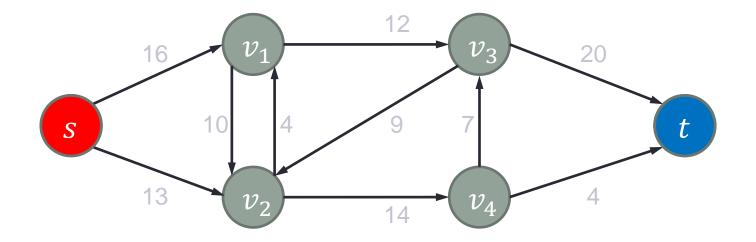


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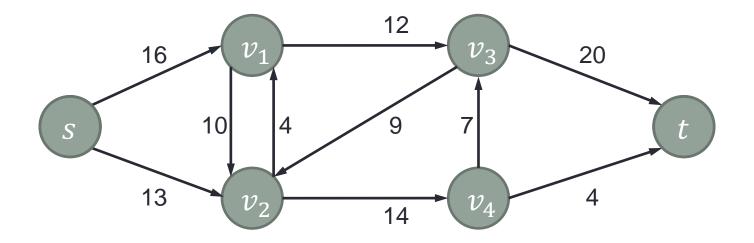


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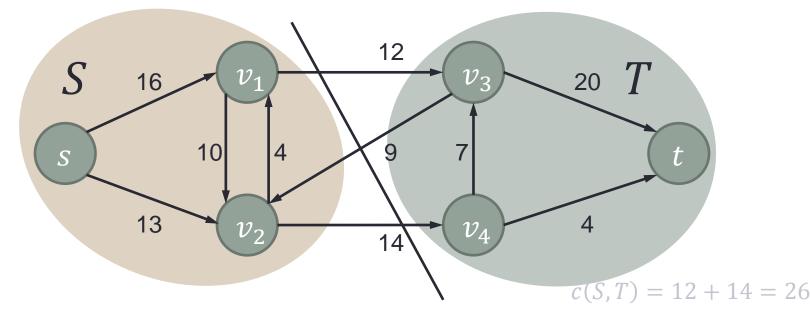
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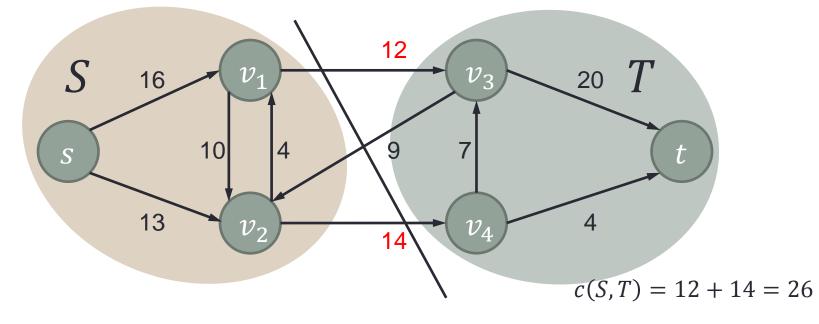
Cut of a Graph

- A cut of the graph shows the division of *V* into two groups; *S* and *T*.
- The capacity of the cut c(S,T) is defined as the capacity of the edges from S to T.
- The cut which minimizes c(S,T) is called min-cut.



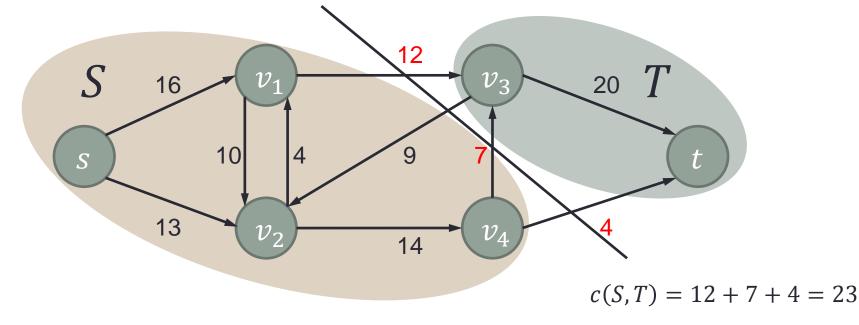
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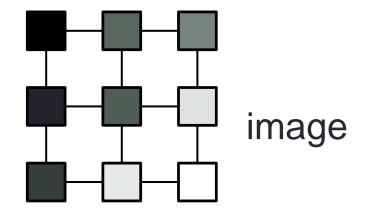
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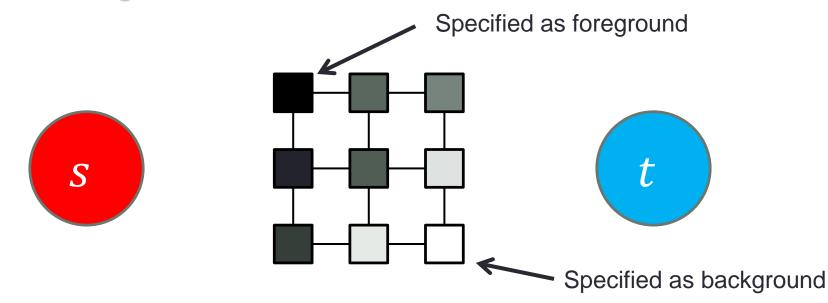
Making a Graph (Color)

 The weighted directed graph is generated from the pixels in the image.



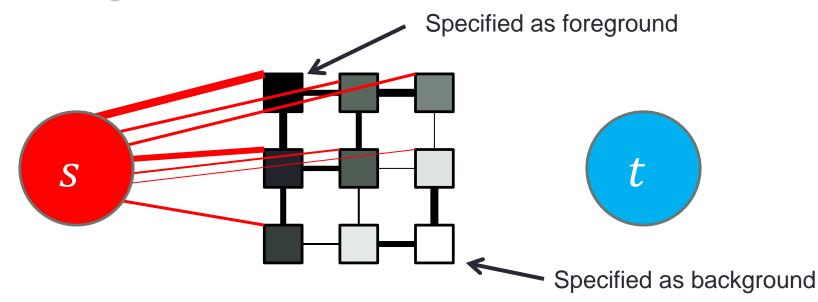
Making a Graph (Seed)

- Suppose that a black pixel is specified as foreground, and a white pixel is specified as background.
- Then, pixels that have similar color to black have strong connection to *s*.
- On the other hand, pixels that have similar color to white have strong connection to *t*.



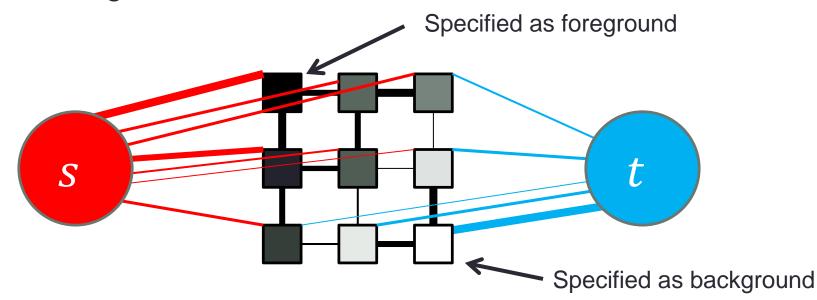
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Making a Graph (Energy)

Min-cut corresponds to the minimum energy of the following equation.

$$E(\mathbf{L}) = \lambda \sum_{p \in V} R_p(L_p) + \sum_{\{p,q\} \in E} B_{\{p,q\}} \cdot \delta(L_p, L_q)$$

 λ is a parameter which controls the affect by the seeds (the larger the value, the more affect by the seeds).

How to Compute Min-cut

- According to "max-flow min-cut theorem", min-cut is obtained from the result of max-flow.
- In order to calculate max-flow, two methods are commonly used.
 - "augmenting path method" scans the graph to find a path from source (s) to sink (t).
 This method is NOT suitable for hardware implementation.
 - 2. "push-relabel method" uses only the connection from one vertex to its neighbors.

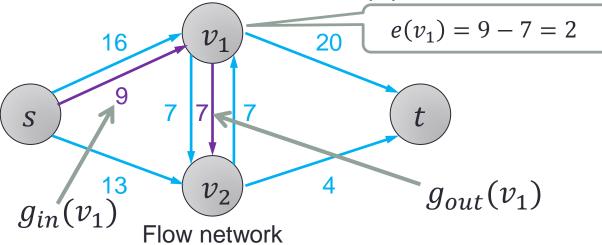
This method is suitable for hardware implementation.

Push-relabel Method

- In the push-relabel method, a weighted directed graph is considered as a flow network.
- We can flow preflow g in each edge if g is smaller than flow capacity c(u, v).
- All vertices have excess flow

$$e(u) = g_{in}(u) - g_{out}(u) \ge 0.$$

• Vertex u is active if e(u) > 0.



Residual Network

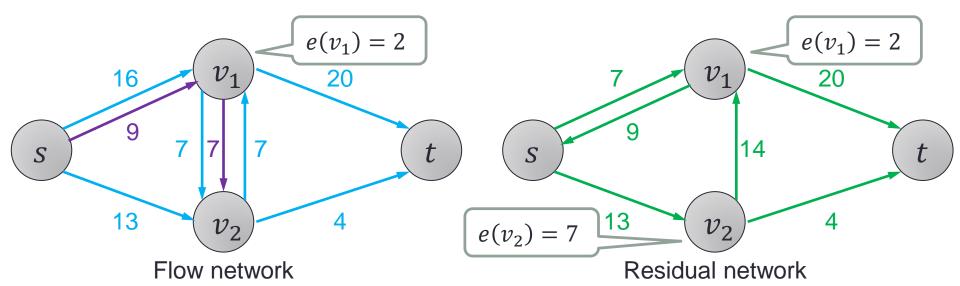
• The residual capacity of an edge is given by $c_f(u, v) = c(u, v) - g(u, v)$

which is the rest of the capacity that we can flow from u to v.

• By flowing 7 from v_1 to v_2 ,

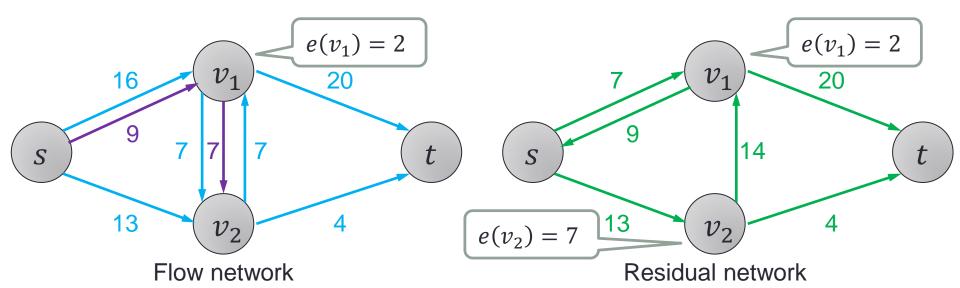
$$c_f(v_1, v_2) = 7 - 7 = 0$$

 $c_f(v_2, v_1) = 7 - (-7) = 14$



Residual Network

- Using the residual network, we can easily understand how much more we can flow on the network.
- However, we must store excess flow of each vertex.

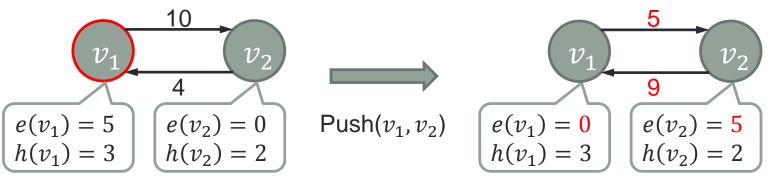


Operations of Push-relabel Method

- There are two main operations, and they are applied to the active vertices.
 - 1. Push(u, v)
 - 2. Relabel(u)
- If u is active, either operation can be applied to u.

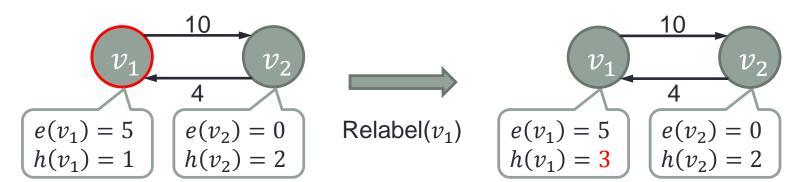
Push(u, v)

- Applicable condition
 - Vertex *u* is active.
 - $c_f(u,v) > 0$
 - h(u) = h(v) + 1
- Operation
 - $\min(e(u), c_f(u, v))$ is flowed from u to v.
- Example
 - Preflow 5 is flowed from v_1 to v_2 .
 - Residual capacity $c_f(v_1, v_2)$ is reduced, and $c_f(v_2, v_1)$ is increased.



Relabel(u)

- Applicable condition
 - Vertex *u* is active.
 - Push(u, v) cannot be applied to vertex u.
- Operation
 - h(u) is heightened so that push(u, v) can be applied.
- Example
 - Push (v_1, v_2) can not applied to v_1 because $h(v_1) < h(v_2)$.
 - h(v₁) is heightened more than h(v₂) so that push(v₁, v₂) can be applied to v₁.



Heuristics for the Push-relabel Method

- The computational complexity of the push-relabel method is $O(V^2E)$.
- To reduce the computational complexity, two heuristics are widely used.
- "global relabeling" changes h(u) by calculating the minimum distance from u to t by the breadth first search.
 We need to traverse the graph by dereferencing, so it is NOT suitable for hardware implementation.
- "gap relabeling" heightens h(u) to |V| + 1 if u belongs to S. This method can be implemented using a histogram.

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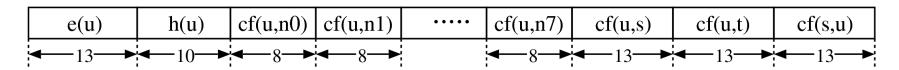
Main Features in Our Approach

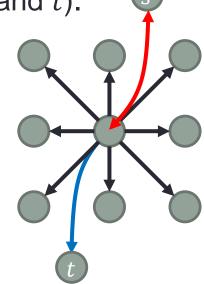
- Major operations are "push" and "ralabel".
- Operations are applied to the active vertices.
- Relabel is applied first if necessary, and then push is applied.
- A FIFO is used to manage active vertices, because the order of the processing is arbitrary.
- We can obtain max-flow of the flow network when there exists no active vertex.

Hardware Implementation

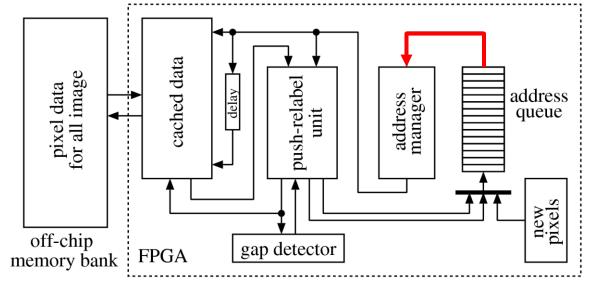
Data Format of Each Pixel

- Each vertex u has 10 links (eight neighbors, and s and t).
- Each link has residual capacity $c_f(u,*)$ from u.
- Vertex u also has residual capacity c_f(s, u) from s to u.
- Excess flow e(u) and height h(u) are required for each vertex u.
- The total data width is 126b.

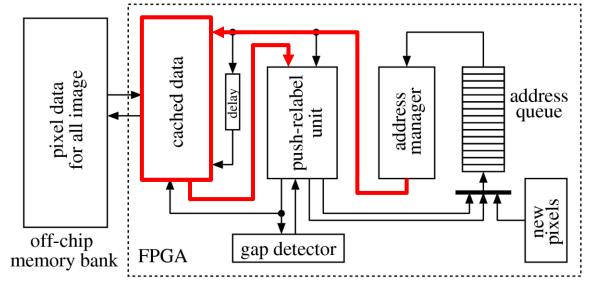




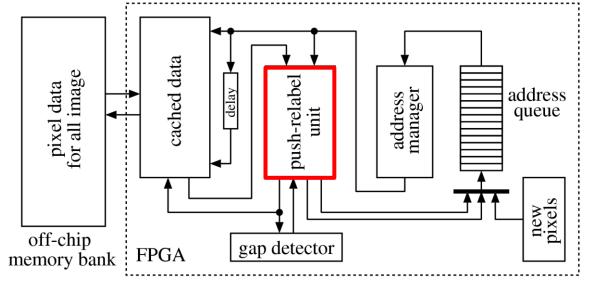
- 1. The address of an active vertex is popped up from the address queue.
- 2. The data of the nine pixels are read out from the cache memory.
- 3. Relabel operation is applied if necessary, and push operation is applied in the push-relabel unit.
- 4. If new active vertex is generated, put it in the address queue.
- 5. The result is written back to the cache memory.



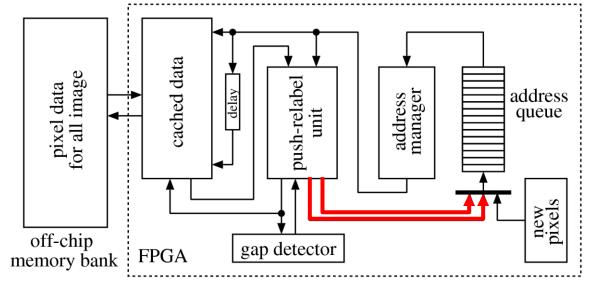
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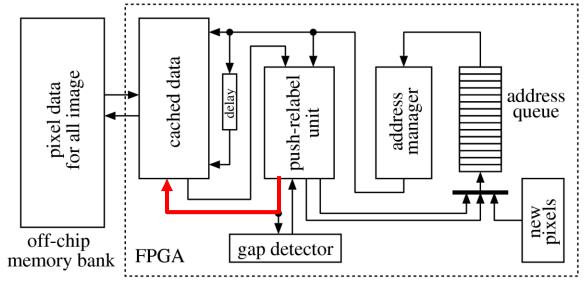
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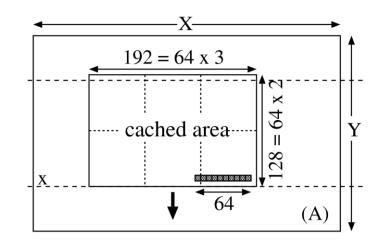
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Data Caching Method

- 192 x 128 pixels are cached on block RAMs.
- The cached area is changed.
- Among the cached pixels, 64 pixels are newly processed.

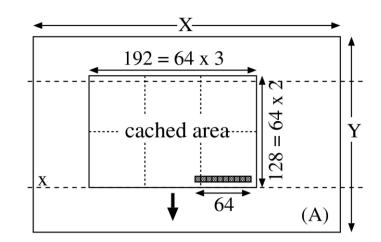




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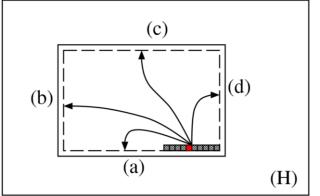




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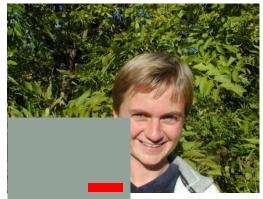
- By applying the push operation to an active pixel, its neighbor pixels may become active from one to another.
- There are four possibilities that active pixels go out of the cached area.
 - In case of (a), those pixels are pushed in a queue and processed afterward.
 - In case of (b) or (c), a control flag is set, and vertical scan is rewound.
 - In case of (d), those processed in the next vertical scan.

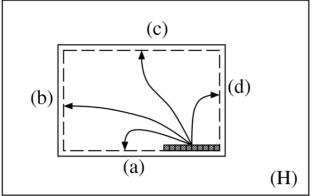




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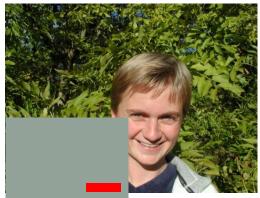
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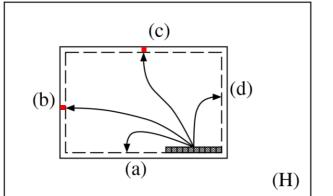




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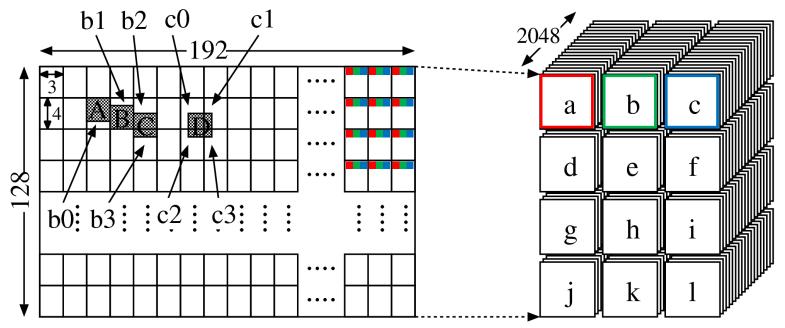
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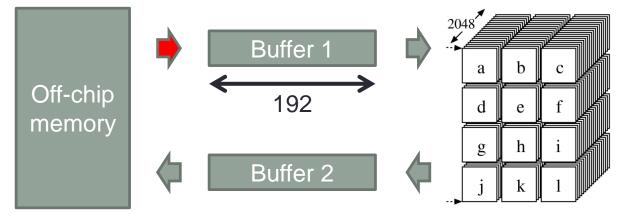
Data Mapping Method

- 192 x 128 pixels in the target area are mapped onto
 12 banks (arranged 3 x 4) to allow parallel accesses to them.
- 12 pixels around any coordinate can be read out in parallel.
- 9 of the 12 pixels are selected by the selectors, and given to the push-relabel unit.



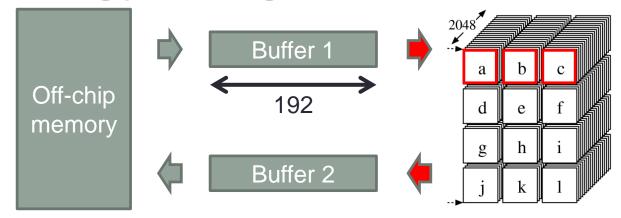
Updating the Cached Area

- The data of the next 192 pixels are read into a set of buffers which consist of distributed RAMs, while the pixels are being processing.
- When the number of active pixels becomes less than the given threshold, the push-relabel unit is stopped.
- The data of 3 of the 12 banks are updated in parallel.
- Old data are written back to the off-chip memory, while the pixels are being processing.



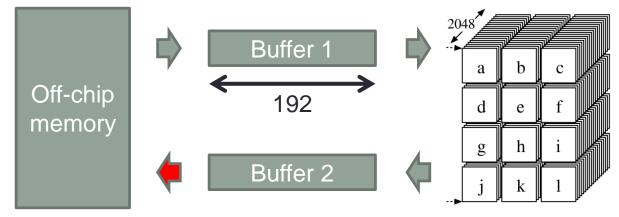
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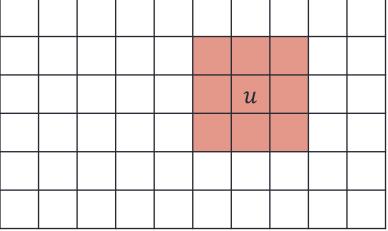


Filling the Pipeline Stages

- Push-relabel unit has 10 stages.
- In order to achieve higher performance, we need to fulfill all the pipeline stages.
- However, while a pixel u is being processed in this unit, its neighbor pixel v can not be put into the unit, because c_f(v,u) may be changed by the processing of u.

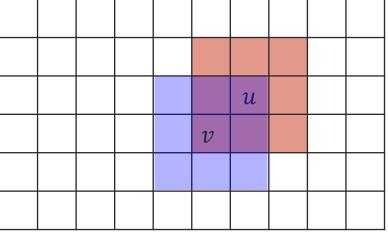
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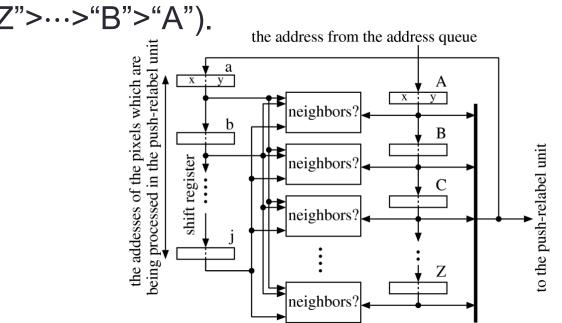
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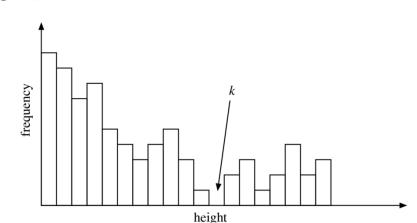
Management of the Pixels

- Suppose that pixels "a" to "j" are being processed.
- First, new pixels are put into the shift register "A" to "Z".
- If one of "a" to "j' is a neighbor of the pixel on "C", the data on "C" continues to stay on the shift register.
- If several pixels can be processed, the older one is chosen (piriority is "Z">···>"B">"A").



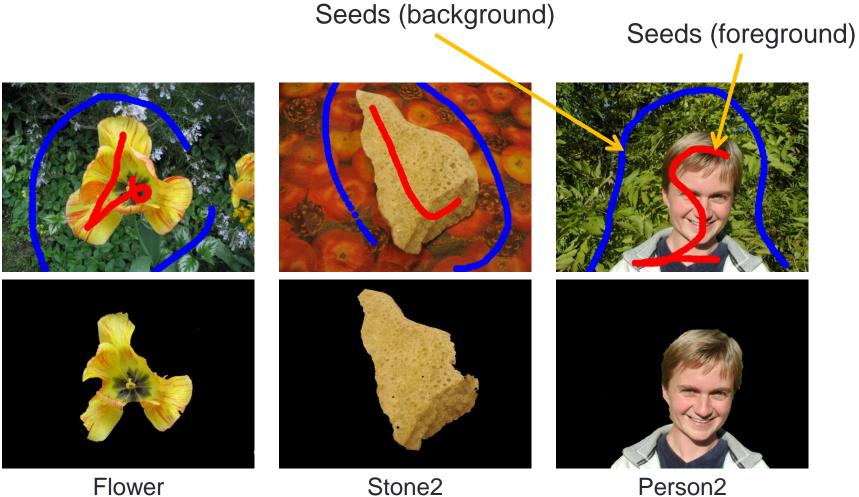
Detecting Gaps Using a Histogram

- h(u) is heightened to |V| + 1 by gap relabeling heuristics if vertex of height = k does NOT exist and h(u) > k.
- In our implementation, k is looked up using a histogram of the height of all pixels.
- In our experiments,
 - The maximum value of each bin is less than 10000, so the data width of counters is 15b.
 - The maximum k is less than 512.
 We used 800 instead of |V|. ↑



Experimental Results

- We have implemented the circuit on Xilinx XC6VLX130T-3.
- The circuit uses 33.3 KLUTs (41%) and 97 36Kb block RAMs (36%).
- Operational frequency is 201.1 MHz.
- We have compared the performance with
 - Software program (maxflow-v3.01) on Intel Core 2 Duo E8500 @3.16 GHz.
 - GPU program on GeForce GTX280.
- The graph is generated on the host computer.

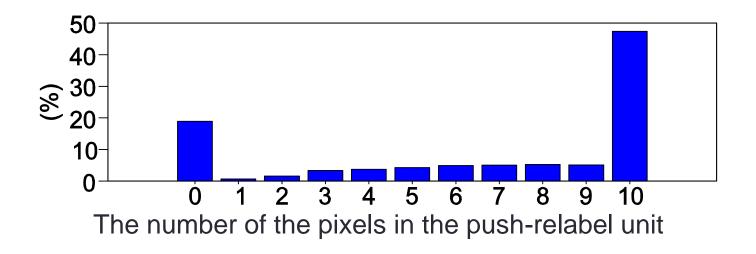


Stone2 Segmentation results

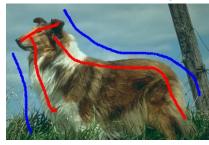
- The performance of the FPGA is almost comparable with GPU (20 -- 30 fps).
- Proposed system is about 3 to 5 times faster than CPU.

Benchmark		Ex	Speedup		
Image	Size	CPU	GPU	FPGA	
Flower	600 x 450	161.1	37	30.7	5.2
Stone2	640 x 480	117.2	44	45.8	2.6
Person2	600 x 450	118.5	61	36.7	3.2

- This figure shows the number of the pixels in the pushrelabel unit when processing Person2.
- All stages are fully filled during about 50% of the execution time.
- But the idle time occupies about 20%.



- Four different seeds are given to "dog".
- The speedup depends on the seeds, but fast enough for real-time processing.

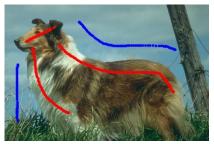


Davida a ala



 $D = \frac{1}{2} \left(\frac{1}{2} - \frac{1}{2} \right)$





Dog / seed1	Dog / se	ed2 Dog	g / seed3	Dog / seed4
Benchmark		Exec. Tiı	me (msec)	Speedup
Image	Size	CPU	FPGA	
Dog / seed1	482 x 321	182.1	31.7	5.7
Dog / seed2	482 x 321	300.0	30.3	9.9
Dog / seed3	482 x 321	185.7	26.4	7.0
Dog / seed4	482 x 321	363.4	33.1	11.0

- Images with small object are segmented.
- Worse speedup because of higher idle ratio of the pipeline stages (processing of the background pixels finishes faster than the pixels on the foreground).
- However, it is fast enough for real-time processing.





	Wolf		Sheep		
Benchmark		Exec. Tim	Speedup		
Image	Size	CPU	FPGA		
Wolf	482 x 321	17.8	7.6	2.3	
Sheep	450 x 600	33.6	16.9	2.0	

Conclusions and Future Work

- We have proposed an acceleration method of the maxflow problem with FPGA.
- The performance gain compared with a software library on CPU is about 3 to 5.
- For more speedup,
 - We need to fill the pipeline stage of the push-relabel unit more.
 - Several push-relabel units can be implemented (the size of the unit is small enough).

Thank you for your kind attention