

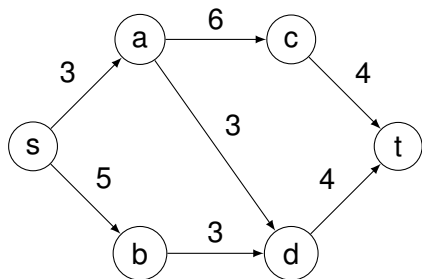
# Formalizing the Edmonds-Karp Algorithm

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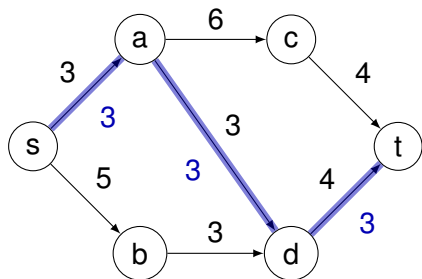
August 2016

# Flow Networks



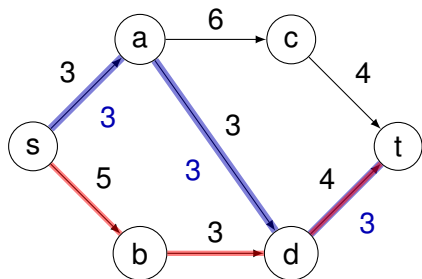
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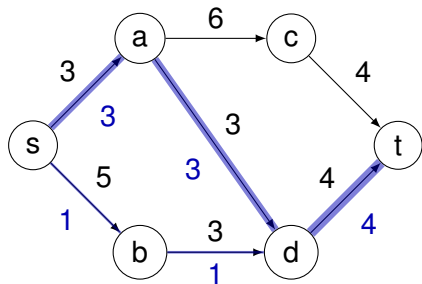
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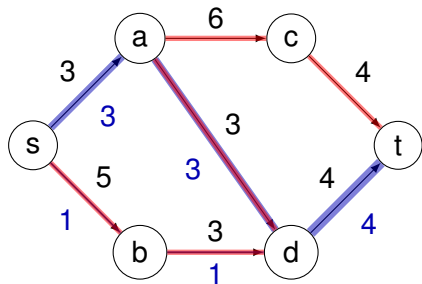
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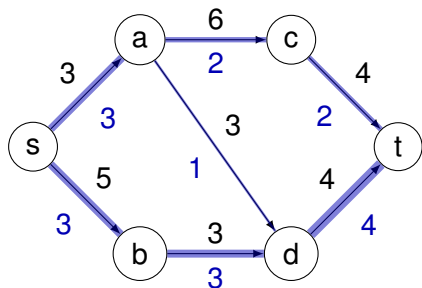
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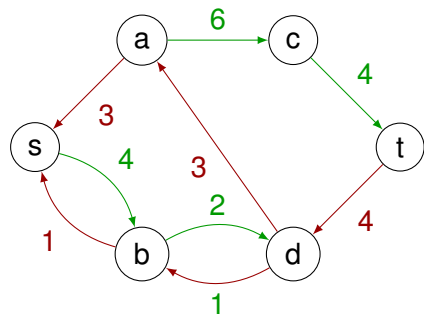


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- Finding maximum flow
  - Incr. along augmenting path
- May need to take back flow
  - To increase overall value
- Flow is maximal now

# Residual Graph

## of Network and Flow

- Flow that can be moved between nodes
  - By **increasing** or **taking back** flow

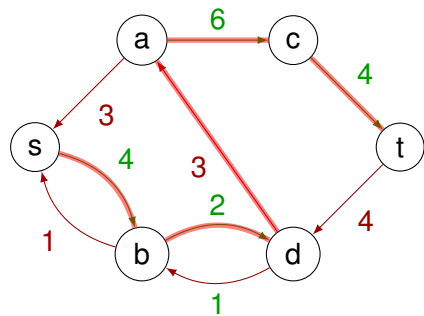




# Residual Graph

## of Network and Flow

- Flow that can be moved between nodes
  - By **increasing** or **taking back** flow
- Augmenting path: s-t path in residual graph



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**while** *exists augmenting path*

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- Partial correctness: obvious
- Termination: only for integer/rational capacities
- Edmonds/Karp: choose shortest augmenting path
  - $O(VE)$  iterations for real-valued capacities
  - Using BFS to find path:  $O(VE^2)$  algorithm

# Our Contributions

Verified in Isabelle/HOL

- Min-Cut/Max-Flow Theorem
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- Ford-Fulkerson and Edmonds Karp algorithms
  - Human-readable presentation of algorithms
  - Proved correctness and complexity
- Efficient Implementation
  - Using stepwise refinement down to Imperative/HOL
  - Isabelle's code generator exports to SML
  - Benchmark: comparable to Java (from Sedgewick et al.)

# Human-Readable Proofs

- Used Isar proof language

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Proof fragment from Cormen et al.:

$$\begin{aligned}(f \uparrow f')(u, v) &= f(u, v) + f'(u, v) - f'(v, u) && \text{(definition of } \uparrow \text{)} \\ &\leq f(u, v) + f'(u, v) && \text{(because flows are nonnegative)} \\ &\leq f(u, v) + c_f(u, v) && \text{(capacity constraint)} \\ &= f(u, v) + c(u, v) - f(u, v) && \text{(definition of } c_f \text{)} \\ &= c(u, v).\end{aligned}$$

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Our Isar version:

```
have (f↑f')(u,v) = f(u,v) + f'(u,v) - f'(v,u)
by (auto simp: augment_def)
also have ... ≤ f(u,v) + f'(u,v) using f.capacity_const by auto
also have ... ≤ f(u,v) + cf(u,v) using f.capacity_const by auto
also have ... = f(u,v) + c(u,v) - f(u,v)
by (auto simp: residualGraph_def)
also have ... = c(u,v) by auto
finally show (f↑f')(u, v) ≤ c(u, v) .
```

# And Automatic Proofs

- Cormen et al. also give more complicated proofs

First part of proof that  $|f \uparrow f'| = |f| + |f'|$ :

$$\begin{aligned} & |f \uparrow f'| \\ &= \sum_{v \in V_1} (f(s, v) + f'(s, v) - f'(v, s)) - \sum_{v \in V_2} (f(v, s) + f'(v, s) - f'(s, v)) \\ &= \sum_{v \in V_1} f(s, v) + \sum_{v \in V_1} f'(s, v) - \sum_{v \in V_1} f'(v, s) \\ &\quad - \sum_{v \in V_2} f(v, s) - \sum_{v \in V_2} f'(v, s) + \sum_{v \in V_2} f'(s, v) \\ &= \sum_{v \in V_1} f(s, v) - \sum_{v \in V_2} f(v, s) \\ &\quad + \sum_{v \in V_1} f'(s, v) + \sum_{v \in V_2} f'(s, v) - \sum_{v \in V_1} f'(v, s) - \sum_{v \in V_2} f'(v, s) \\ &= \sum_{v \in V_1} f(s, v) - \sum_{v \in V_2} f(v, s) + \sum_{v \in V_1 \cup V_2} f'(s, v) - \sum_{v \in V_1 \cup V_2} f'(v, s). \quad (26.6) \end{aligned}$$

## And Automatic Proofs

- Cormen et al. also give more complicated proofs
- We sometimes chose to use more automatic proofs

**lemma** augment\_flow\_value: Flow.val c s (f↑f') = val + Flow.val cf s f'

**proof** -

**interpret** f'': Flow c s t f↑f' **using** augment\_flow\_presv[OF assms] .

## And Automatic Proofs

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- We sometimes chose to use more automatic proofs
  - Using some simplifier setup

**note** `setsum_simp_setup[simp] =`  
`sum_outgoing_alt[OF capacity_const] s_node`  
`sum_incoming_alt[OF capacity_const]`  
`cf.sum_outgoing_alt[OF f'.capacity_const]`  
`cf.sum_incoming_alt[OF f'.capacity_const]`  
`sum_outgoing_alt[OF f''.capacity_const]`  
`sum_incoming_alt[OF f''.capacity_const]`  
`setsum_subtractf setsum.distrib`

## And Automatic Proofs

- Cormen et al. also give more complicated proofs
- We sometimes chose to use more automatic proofs
  - Using some simplifier setup
  - And auxiliary statements

**have** aux1:  $f'(u,v) = 0$  **if**  $(u,v) \notin E$   $(v,u) \notin E$  **for**  $u\ v$

**proof** -

**from** that  $cfE\_ss\_invE$  **have**  $(u,v) \notin cf.E$  **by** auto

**thus**  $f'(u,v) = 0$  **by** auto

**qed**



## And Automatic Proofs

- Cormen et al. also give more complicated proofs
- We sometimes chose to use more automatic proofs
  - Using some simplifier setup
  - And auxiliary statements
  - We reduce the displayed proof's complexity

**have**  $f''.val = (\sum u \in V. \text{augment } f' (s, u) - \text{augment } f' (u, s))$

**unfolding**  $f''.val\_def$  **by** `simp`

**also have**  $\dots = (\sum u \in V. f (s, u) - f (u, s) + (f' (s, u) - f' (u, s)))$

— Note that this is the crucial step of the proof, which Cormen et al. leave as an exercise.

**by** `(rule setsum.cong) (auto simp: augment_def no_parallel_edge aux1)`

**also have**  $\dots = val + \text{Flow.val } cf \ s \ f'$

**unfolding**  $val\_def \ f'.val\_def$  **by** `simp`

**finally show**  $f''.val = val + f'.val$  .

**qed**

# Main Result

- Finally, we arrive at

**context** NFlow **begin**

...

**theorem** ford\_fulkerson:

isMaxFlow  $f \iff (\nexists p. \text{isAugmentingPath } p)$

# Ford-Fulkerson Method

- We use the Isabelle Refinement Framework

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```
definition ford_fulkerson_method  $\equiv$  do {  
  let  $f_0 = (\lambda(u,v). 0)$ ;  
  
  (f,brk)  $\leftarrow$  while ( $\lambda(f,brk). \neg brk$ )  
  ( $\lambda(f,brk). \mathbf{do}$  {  
    p  $\leftarrow$  selectp p. is_augmenting_path f p;  
    case p of  
      None  $\Rightarrow$  return (f,True)  
    | Some p  $\Rightarrow$  return (augment c f p, False)  
  })  
  (f0,False);  
  return f  
}
```

# Correctness Proof

- First, we add some assertions and invariant annotations

```
definition fofu  $\equiv$  do {  
  let f0 = ( $\lambda$ _. 0);  
  
  (f,_)  $\leftarrow$  while fofu_invar  
  ( $\lambda$ (f,brk).  $\neg$ brk)  
  ( $\lambda$ (f,_) do {  
    p  $\leftarrow$  find_augmenting_spec f;  
    case p of  
      None  $\Rightarrow$  return (f,True)  
    | Some p  $\Rightarrow$  do {  
      assert (p $\neq$ []);  
      assert (NFlow.isAugmentingPath c s t f p);  
      let f = NFlow.augment_with_path c f p;  
      assert (NFlow c s t f);  
      return (f, False)  
    }  
  })  
  (f0,False);  
  assert (NFlow c s t f);  
  return f  
}
```

# Correctness Proof

- First, we add some assertions and invariant annotations
- Then, we use the VCG to prove partial correctness

**theorem** fofu\_partial\_correct: fofu  $\leq$  (**spec** f. isMaxFlow f)

**unfolding** fofu\_def find\_augmenting\_spec\_def

**apply** (refine\_vcg)

**apply** (vc\_solve simp:

zero\_flow

NFlow.augment\_pres\_nflow

NFlow.augmenting\_path\_not\_empty

NFlow.noAugPath\_iff\_maxFlow[symmetric])

**done**

## Correctness Proof

- First, we add some assertions and invariant annotations
- Then, we use the VCG to prove partial correctness
- This also yields correctness of the unannotated version

**theorem** (in Network) ford\_fulkerson\_method  $\leq$  (**spec** f. isMaxFlow f)



# Edmonds-Karp Algorithm

- Specify shortest augmenting path

**definition** find\_shortest\_augmenting\_spec f  $\equiv$

**assert** (NFlow c s t f)  $\gg$

(**selectp** p. Graph.isShortestPath (residualGraph c f) s p t)

# Edmonds-Karp Algorithm

- Specify shortest augmenting path
- This is a refinement of augmenting path

**lemma** `find_shortest_augmenting_refine`:

`find_shortest_augmenting_spec`  $\leq$  `find_augmenting_spec`

# Edmonds-Karp Algorithm

- Specify shortest augmenting path
- This is a refinement of augmenting path
- Replace in algorithm

**definition** `fofu`  $\equiv$  **do** {

...

`p`  $\leftarrow$  `find_augmenting_spec f`;

...

# Edmonds-Karp Algorithm

- Specify shortest augmenting path
- This is a refinement of augmenting path
- Replace in algorithm

**definition** `edka_partial`  $\equiv$  **do** {

...

`p`  $\leftarrow$  `find_shortest_augmenting_spec` `f`;

...

# Edmonds-Karp Algorithm

- Specify shortest augmenting path
- This is a refinement of augmenting path
- Replace in algorithm
- New algorithm refines original one

**lemma** edka\_partial\_refine[refine]: edka\_partial  $\leq$  fofu

**unfolding** find\_shortest\_augmenting\_spec\_def find\_augmenting\_spec\_def  
**apply** (refine\_vcg)  
**apply** (auto  
  simp: NFlow.shortest\_is\_augmenting  
  dest: NFlow.augmenting\_path\_imp\_shortest)  
**done**

# Total Correctness and Complexity

- Next, we define a total correct version

**definition** `edka_partial`  $\equiv$  **do** {

...

`(f,_)`  $\leftarrow$  **while**<sup>*fofu\_invar*</sup>

...

# Total Correctness and Complexity

- Next, we define a total correct version

**definition**  $edka \equiv \mathbf{do} \{$

...

$(f, \_) \leftarrow \mathbf{while}_T^{f \text{ of } u \text{ } \_ \text{ } invar}$

...

## Total Correctness and Complexity

- Next, we define a total correct version
- And show refinement

**theorem** `edka_refine[refine]`:  $\text{edka} \leq \text{edka\_partial}$



# Total Correctness and Complexity

- Next, we define a total correct version
- And show refinement
- We also show  $O(VE)$  bound on loop iterations
  - Instrumenting the loop with a counter

# Towards Efficient Implementation

Several refinement steps lead to final implementation:

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  - Tabulate capacity matrix and adjacency map to array
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  - Tabulate capacity matrix and adjacency map to array
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- 6 Export to SML code



# Assembling Overall Correctness Proof

- Correctness statement
  - As Hoare Triple using Separation Logic

**context** Network\_Impl **begin**

**theorem** edka\_imp\_correct:

**assumes** Graph.V c  $\subseteq$  {0.. $N$ }

**assumes** is\_adj\_map am

**shows**

<emp>

edka\_imp c s t N am

< $\lambda$ fi.  $\exists_A$  f. is\_rflow N f fi \*  $\uparrow$ (isMaxFlow f)><sub>t</sub>

# Assembling Overall Correctness Proof

- Correctness statement
  - As Hoare Triple using Separation Logic
- Proof by transitivity

**proof -**

**interpret** Edka\_Impl **by** unfold\_locales fact

**note** edka5\_refine[OF ABS\_PS]

**also note** edka4\_refine

**also note** edka3\_refine

**also note** edka2\_refine

**also note** edka\_refine

**also note** edka\_partial\_refine

**also note** fofu\_partial\_correct

**finally have** edka5 am  $\leq$  SPEC isMaxFlow .

**from** hn\_refine\_ref[OF this edka\_imp\_refine]

**show** ?thesis

**by** (simp add: hn\_refine\_def)

**qed**

# Assembling Overall Correctness Proof

- Correctness statement
  - As Hoare Triple using Separation Logic
- Proof by transitivity
- Also integrated with check for valid network
  - Input: list of edges, source node, sink node

## theorem

**fixes** el **defines** c  $\equiv$  In<sub>α</sub> el

**shows**

<emp>

edmonds\_karp el s t

<

λNone  $\Rightarrow$   $\uparrow(\neg \text{In\_invar } el \vee \neg \text{Network } c \ s \ t)$

| Some (l,\_,N,cf)  $\Rightarrow$

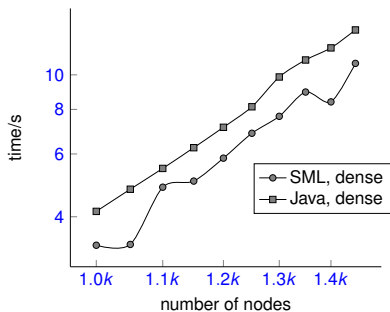
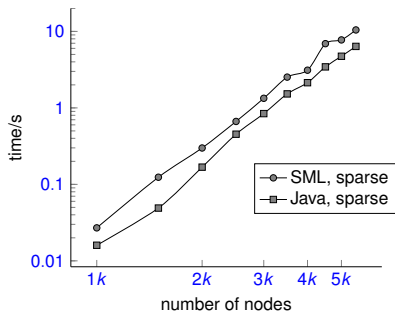
$\uparrow(\text{In\_invar } el \wedge \text{Network } c \ s \ t \wedge \text{Graph.V } c \subseteq \{0..<N\})$

\*  $(\exists_A f. \text{is\_rflow } c \ s \ t \ N \ f \ cf \ * \uparrow(\text{Network.isMaxFlow } c \ s \ t \ f))$

><sub>t</sub>

# Benchmarking

- Against Java version of Sedgewick et al., on random networks
- Two data sets: Sparse ( $D = 0.02$ ) and dense ( $D = 0.25$ ) graphs
  - Sparse: Java is (slightly) faster
  - Dense: we are (slightly) faster
  - Supposed reason: different 2-dimensional array implementations



# Conclusion

- Proof of Min-Cut/Max-Flow Theorem
  - Human readable proofs following textbook presentation
  - Showing off Isar proof language
- Verified Edmonds-Karp algorithm
  - From abstract pseudo-code like version ...
  - ... down to imperative implementation
  - Showing off Isabelle Refinement Framework
- Our implementation is pretty efficient

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## Questions?