

Gradient of MPI-parallel codes

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Outline

- 1 Adjoints in a sequential context
- 2 Parallel Adjoints: global communication
- 3 Parallel Adjoints: point-to-point communication
- 4 Remark: waitall
- 5 Data-Dependence graphs: a tool for formal validation ?
- 6 Data-Flow analyses in AD tools
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- 8 Conclusion and Outlook

Adjoint Algorithms

Given an **algorithm** P that computes a **function** f ,
the **“adjoint algorithm”** \bar{P} computes the **gradient** of f .

Adjoint Algorithms

- compute the gradient at a cost that is **independent** from the number of inputs of f .
- compute the gradient backwards
⇒ **reversal** of the control-flow and of the data-flow.
- can be built from P by an **Automatic Differentiation** tool.

Structure of Adjoint Programs

Algorithm P:

input a, b, c

...

...

$u = g(a, c)$

...

...

$r = h(u, v)$

output r

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Algorithm \bar{P} :

input \bar{r}

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Algorithm \bar{P} :

$\bar{r} = 0.0$

$\bar{v} += \frac{\partial h}{\partial v} * \bar{r}$

$\bar{u} += \frac{\partial h}{\partial u} * \bar{r}$

input \bar{r}

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Algorithm \bar{P} :

$$\bar{u} = 0.0$$

$$\bar{c} += \frac{\partial g}{\partial c} * \bar{u}$$

$$\bar{a} += \frac{\partial g}{\partial a} * \bar{u}$$

...

...

$$\bar{r} = 0.0$$

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output $\bar{a}, \bar{b}, \bar{c}$

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Adjoining Simple Instructions

Original:	Adjoint:
$x = a + 2*b$	$\bar{a} = \bar{a} + \bar{x} ; \bar{b} = \bar{b} + 2*\bar{x} ; \bar{x} = 0$
$x = 2*x$	$\bar{x} = 2*\bar{x}$
$y = \sin(x)$	$\bar{x} = \bar{x} + \cos(x)*\bar{y} ; \bar{y} = 0$
$b = a$	$\bar{a} = \bar{a} + \bar{b} ; \bar{b} = 0$
$s = \text{SUM}(T(:))$	$\bar{T}(:) = \bar{T}(:) + \bar{s} ; \bar{s} = 0$
$U(2:9) = U(2:9) + x$	$\bar{x} = \bar{x} + \text{SUM}(\bar{U}(2:9))$
$\text{where}(T>3) T = T - a$	$\bar{a} = \bar{a} - \text{SUM}(\bar{T}, T>3)$

All these can be “proved” formally ... but a convenient justification is **backwards propagation of the influence** on the result.

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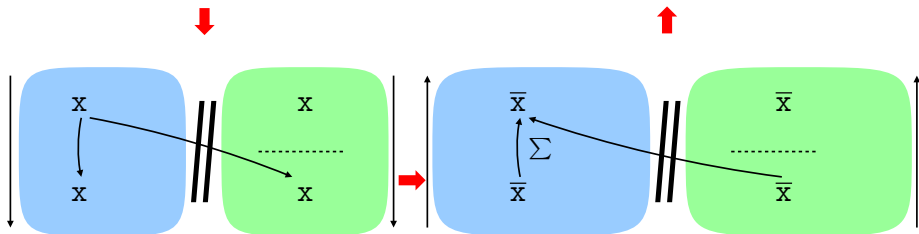
Adjoining one Global Communication

`bcast(x, ,P,)`

`reduce($\bar{x}, \bar{t}, , , \text{SUM}, P,)$`

$\bar{x} = 0.0$

on P: $\bar{x} = \bar{x} + \bar{t}$



Adjoining Global Communications

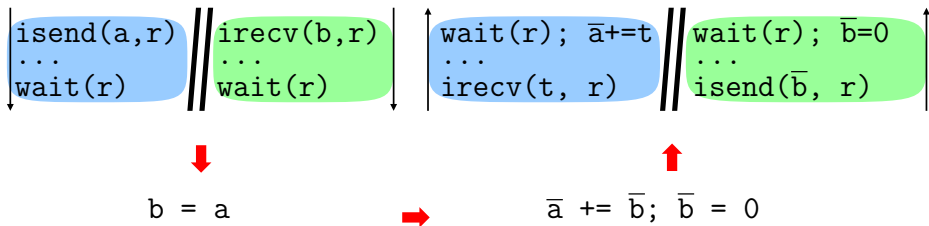
Original:	Adjoint:
<code>bcast(x, ,P,)</code>	<code>reduce(\bar{x}, \bar{t}, , ,SUM,P,)</code> <code>$\bar{x}=0.0$; on P: $\bar{x}=\bar{x}+\bar{t}$</code>
<code>reduce(x,y, , ,SUM,P,)</code>	<code>on P: $\bar{t}=\bar{y}$; on P: $\bar{y}=0.0$</code> <code>bcast(\bar{t}, ,P,)</code> ; <code>$\bar{x}=\bar{x}+\bar{t}$</code>
<code>allreduce(x,y, , ,SUM,)</code>	<code>allreduce(\bar{y}, \bar{t}, , ,SUM,)</code> <code>$\bar{y}=0.0$; $\bar{x}=\bar{x}+\bar{t}$</code>
<code>gather(x, ,y, ,P,)</code>	<code>scatter(\bar{y}, ,\bar{t},P,)</code> <code>on P: $\bar{y}=0.0$; $\bar{x}=\bar{x}+\bar{t}$</code>
<code>scatter(x, ,y, ,P,)</code>	<code>gather(\bar{y}, ,$\bar{t}(:)$, ,P,)</code> <code>$\bar{y}=0.0$; on P: $\bar{x}=\bar{x}+\bar{t}(:)$</code>

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Adjoining Point-to-point Communication

- Blocking \Leftrightarrow {Nonblocking ; wait }
- Analogy `send(a)/receive(b)` with `b = a`



Gives us the adjoints of MPI `isend`, `irecv`, and `wait`.

Rules for adjoining MPI calls

Adjoining rules depend on the context

- between nonblocking calls and their wait
- between the two communicating processes.

in P		in \bar{P}
call	paired with	call
<code>isend(a,r)</code>	<code>wait(r)</code>	<code>wait(r); $\bar{a}+=t$</code>
<code>wait(r)</code>	<code>isend(a,r)</code>	<code>irecv(t,r)</code>
<code>irecv(b,r)</code>	<code>wait(r)</code>	<code>wait(r); $\bar{b}=0$</code>
<code>wait(r)</code>	<code>irecv(b,r)</code>	<code>isend(\bar{b},r)</code>
<code>bSEND(a)</code>	<code>recv(b)</code>	<code>recv(t); $\bar{a}+=t$</code>
<code>recv(b)</code>	<code>bSEND(a)</code>	<code>bSEND(\bar{b}); $\bar{b}=0$</code>
<code>ssend(a)</code>	<code>recv(b)</code>	<code>recv(t); $\bar{a}+=t$</code>
<code>recv(b)</code>	<code>ssend(a)</code>	<code>ssend(\bar{b}); $\bar{b}=0$</code>

Adjoinable MPI requires more information

Adjoining a `wait` requires some context:

- Each `wait` must know what it waits for (`isend` or `irecv`)
- Each `wait` must know and **see** the travelling variable

Source analysis could find matching `wait` of a non-blocking call, but

- not always, and
- making the travelling variable visible is harder.

Instead, use an adjoinable MPI such as:

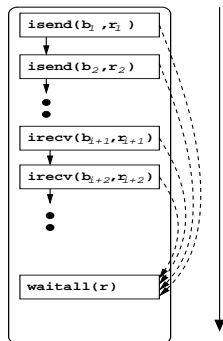
- `waitrecv(b,r)`, whose adjoint is `isend(\bar{b} , r)`
- `waitsend(a,r)`, whose adjoint is `irecv(t, r)`

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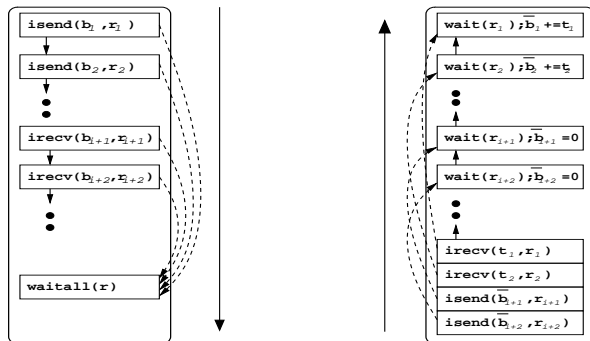
Open question: grouped wait's

“waitall” groups “wait” operations. Improves efficiency.



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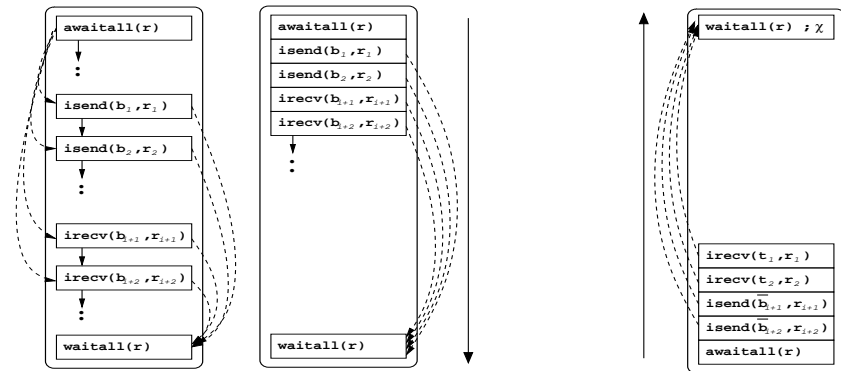
Adjoining re-introduces separate wait's !

⇒ can we get a waitall() back?

The “anti waitall”

Sometimes, we may introduce a `awaitall()`

- nonoperational
- placed by the end-user.



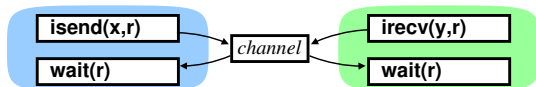
⇒ Allows the adjoint to use a `waitall()` again.

Outline

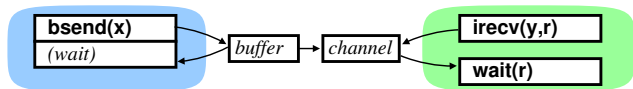
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Data dependence graph for Point-to-Point

- Introduce “*channel*” pseudo variables.
- Nonblocking *isend/irecv*

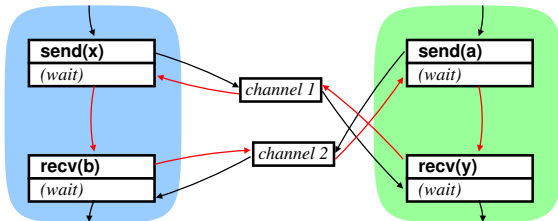


- $\text{send} \Leftrightarrow \{\text{isend}; \text{wait}\}$ $\text{recv} \Leftrightarrow \{\text{irecv}; \text{wait}\}$
- Buffered *bSEND* uses an intermediate copy buffer
⇒ immediate return.



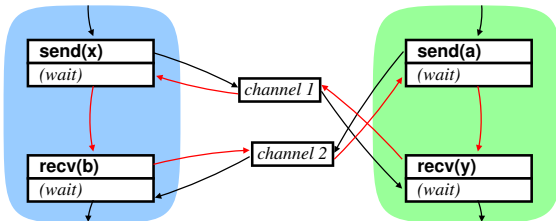
Deadlocks ; Blocking vs Nonblocking

Deadlocks are cycles in the data dependence graph:

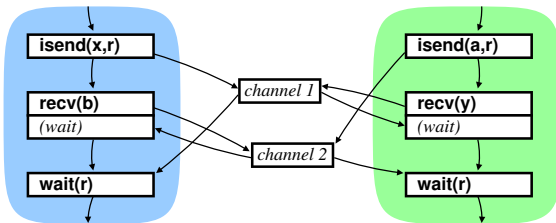


Deadlocks ; Blocking vs Nonblocking

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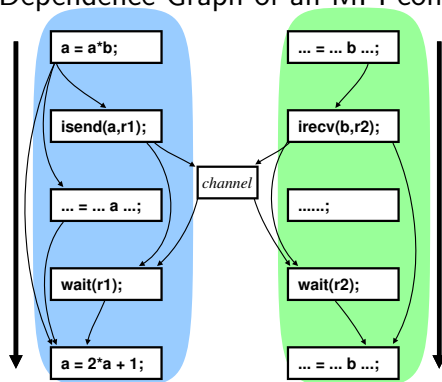
Splitting the wait from the isend/irecv can solve the problem:



Otherwise, use `bSEND`'s.

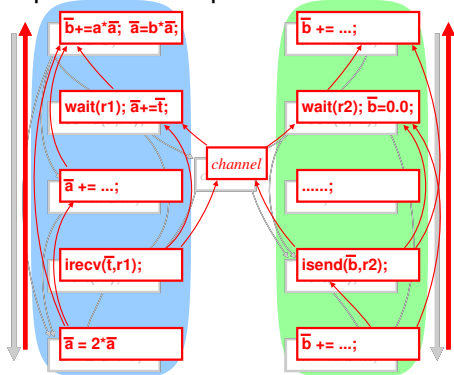
Data Dependence Graph of the Adjoint Algorithm

Consider the Data Dependence Graph of an MPI communication.



Data Dependence Graph of the Adjoint Algorithm

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The Data Dependence Graph of the adjoint communication seems to just reverse the arrows.

⇒ Adjoining does not introduce deadlocks.

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Impact on Data-Flow analyses

AD, like other program transformations, needs preliminary data-flow analyses

Message-passing modifies data-flow \Rightarrow we must adapt analyses.

- If Interprocedural Control-Flow Graph \rightarrow introduce special flow arrows that only convey messages.
- If Call Graph of Flow Graphs \rightarrow introduce special "channel" variables and organize additional fixed-point iterations.

In any case, increases complexity/cost of analyses.

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References

- Utke, Hascoët, Heimbach, Hill, Hovland, Naumann **“Towards Adjoinable MPI”** *IEEE IPDPS, 2009*
- Schanen, Naumann, Hascoët, Utke **“Interpretative Adjoint for Numerical Simulation Codes using MPI”** *Procedia Computer Science, 2010*
- Schanen, Naumann, Hascoët, Utke **“Interpretative Adjoint for Numerical Simulation Codes using MPI”** *ICCS 2010*
- Schanen, Förster, Naumann **“Second-Order Algorithmic Differentiation by Source Transformation of MPI Code”** *EuroMPI 2010*
- Naumann, Hascoët, Hill, Hovland, Riehme, Utke **“A Framework for Proving Correctness of Adjoint Message-Passing Programs”** *PVM/MPI Users’ Group Meeting, 2008*
- Strout, Kreaseck, Hovland **“Data-Flow analysis for MPI programs”** *ICPP, 2006*
- Kreaseck, Strout, Hovland **“Depth Analysis of MPI programs”** *AMP workshop, PLDI 2010*
- Pascual, Hascoët **“Native handling of Message-Passing communication in Data-Flow analysis”** *AD2012*

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Conclusion: an adjoinable MPI ?

- Unlike previous black-box approaches, we consider the level of the **individual MPI calls**.
- Our approach requires the **correspondence** between MPI calls
- Static data-flow analysis will not find it in general:
⇒ **User input** is necessary:
 - could be pragmas
 - could be an “adjoinable MPI” library

- Ongoing **application to the MITgcm**: validates the adjoining rules of MPI calls. Requires the `awaitall`.
- Look for a **general proof** of correctness, maybe based on PGAS or other abstractions e.g. Data Dependence graphs.
- Develop an **adjoinable MPI library**, to help/induce the user write an adjoinable code..

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Thank you for your attention !