An Application of Constraint Programming to the Design and Operation of Synthetic Aperture Radars

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Presentation Outline

- Fragmentation of logical model of SARs
- Constraint programming paradigm and propagation networks
- SAR Inference Engine
- Current areas of investigation



Synthetic Aperture Radars

- Synthetic aperture radars (SARs) image the earth's surface using microwaves.
- SARs are complex systems
 - Hundreds of quantities: center frequency, beamwidth, scene dimensions, etc.
 - Quantities must obey hundreds of relationships: physics, radar equation, trigonometry, etc.



Logical Model of SARs

In aggregate, the quantities and relationships

- Form a large web or network
- Constitute a logical model of the SAR
- Logical model is dispersed
 - People's minds, documents, software
 - Many partially overlapping subsets
- Inconsistencies invariably creep in
 - Cause degraded performance or faults



SAR Inference Engine

- SAR Inference Engine
 - provides a central, common SAR model
 - uses the constraint programming paradigm.
- Constraints come from
 - physics, geometry, signal processing
 - system engineer design choices.
- A propagation network provides the computational foundation.
 - All propagators derived from constraints.



 An imperative code may compute a propagating wave's one way travel time using the following code:

```
velocity = frequency * wavelength; // Equation 1
traveltime = distance / velocity; // Equation 2
```

- Inputs: frequency, wavelength, and distance
- Output: velocity, traveltime



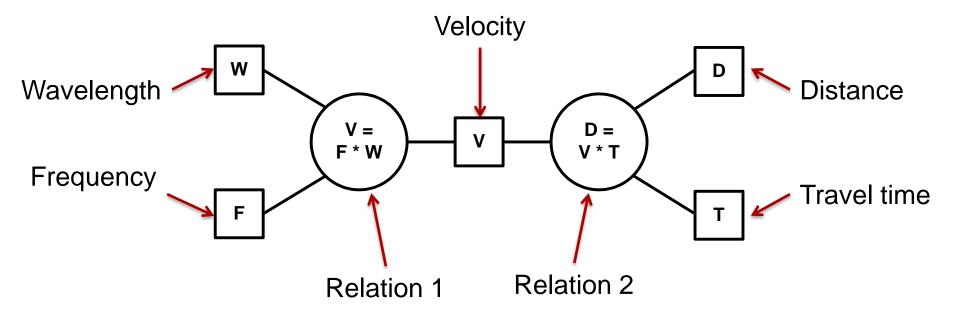
- Alternative arrangements of Equation 1:
 - velocity = frequency * wavelength;
 - wavelength = velocity / frequency;
 - frequency = velocity / wavelength;
- Alternative arrangements of Equation 2:
 - distance = velocity * traveltime;
 - traveltime = distance / velocity;
 - velocity = distance / traveltime;



- Relation 1:
 - velocity = frequency * wavelength
- Relation 2:
 - distance = velocity * traveltime
- Observations:
 - Given any two quantities in a relation the third quantity can be calculated.
 - Both relations have the form "A = B * C" (more on this later).



Relations 1 and 2 depicted as a graph:



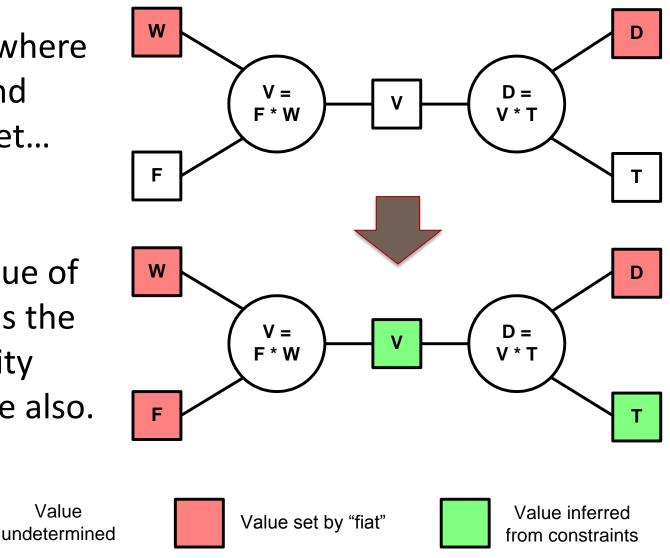
Note: The graph depicts a simple propagation network.



Scenario 1: Determine Travel Time

Given a state where wavelength and distance are set...

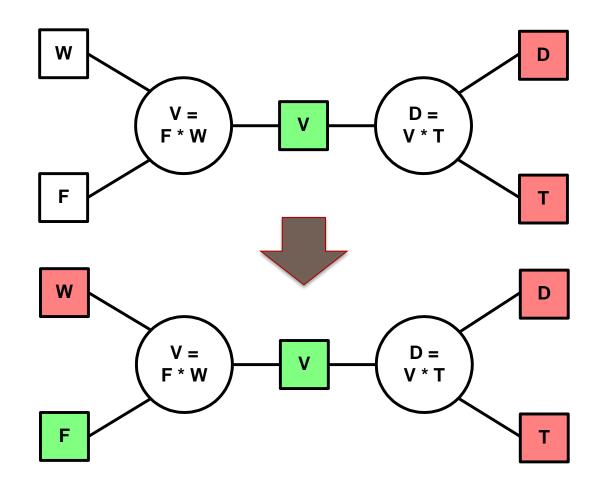
setting the value of frequency fixes the value of velocity and travel time also.



Scenario 2: Determine Frequency

Given a state where distance and travel time are set...

setting the value of wavelength fixes the value of frequency also.





Propagation Network Observations

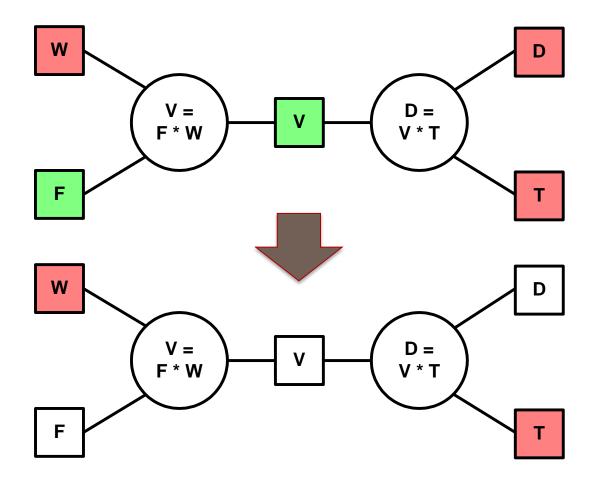
- Many different sets of inputs are possible.
- Relationships can also be inequalities.
- Operation is progressive.
- Useful to record dependencies of "inferred" values on values determined by "fiat"
 - Enables backtracking
 - Enables determining why a quantity is in its current state.



Example of Selective Backtracking

Before invalidating distance

After invalidating distance





Hello World Example

Step 1: Include header files of •Inference Engine •Constraints •Quantities

#include "CenterFrequencyQuantity.h"
#include "CenterWavelengthQuantity.h"
#include "SpeedOfLightQuantity.h"

#include <stdio.h>

#include "InferenceEngine.h"

#include "PropWaveConstraint.h"

Step 2: Create Inference Engine and populate it with constraints

int main (int argc, char * const argv[]) {

// Create Inference Engine and initialize with constraints
InferenceEngine *inferenceEngine = new InferenceEngine ();
inferenceEngine->addConstraint (PropWaveConstraint::getInstance ());
inferenceEngine->concludeInitialization ();

// Assign values to center frequency and speed of light quantities
inferenceEngine->assignQuantityValue (SPEED_OF_LIGHT_NAME, 2.99739141e+008, 0);
inferenceEngine->assignQuantityValue (CENTER_FREQUENCY_NAME, 1.00000000e+008, 0);

// Get value of wavelength
double centerWavelengthValue = Quantity::getInstance (CENTER_WAVELENGTH_NAME)->getValue ();

// Print out result
fprintf (stdout, "CenterWavelength = %.17e \n", centerWavelengthValue);

Step 3b: Extract values of outputs



return 0;

Scaling Up: Comprehensive SAR Model

- Can a SAR be modeled?
- Full-scale SAR model:
 - Port of existing Matlab model that was used to develop two generations of SARs
 - Over 300 Quantities of "physical interest"
 - Over 250 Constraints of "physical interest"
 - Example of diverse quantities included: SNR, geometry, hardware delays, resolution



More on Constraints

- SAR Inference Engine's modeling:
 - Just 11 types of constraints
 - Low-level, i.e. minimal semantic content
 - High-level constraints expressed as multiple lowlevel ones
- Most important types of constraints:
 - A = B op C where "op" is either "+" or "*"
 - "Triangle Constraint" among 3 angles and 3 lengths making up a triangle



Usage Throughout the SAR Lifecycle

Design phase

- Batch mode: generate performance curves
- Interactive mode: explore design space
- Mission planning
 - Handle unanticipated conops
- Radar operation
 - Radar operator interface
 - Embedded in radar



Example: Insertion Into Image Formation

Image Formation Code

Initialization Code:

Among other things, maps attributes of input data onto attributes of output image.

Processing Code:

This code transforms digitized data into SAR images.

Replace mapping logic with calls to SAR Inference Engine.

- Code has two phases
 - Setup / initialization
 - Data processing
- Setup phase maps data attributes to image attributes.
- Mapping process intimately tied to logical model of SAR.
 - Generation of data involved the inverse mapping.
- Machine generated code solves speed issue.



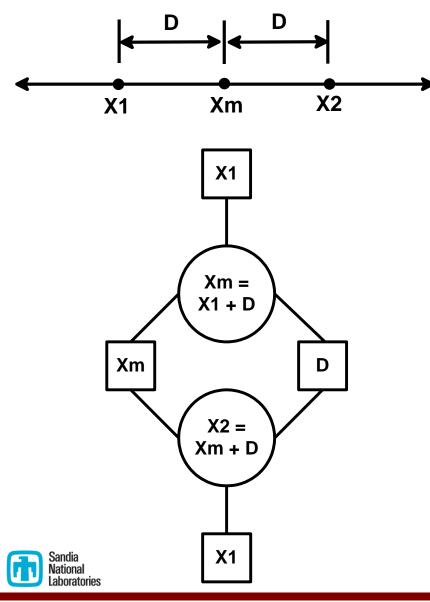
Current Areas of Investigation

- Multivalued solutions
 - E.g. ambiguous case of Law of Sines
- Completeness: Are there cases where inferences could be made but aren't?
 - E.g. set of triangles to fully define geometry
- Constraint Satisfaction Problem issues:
 - Constraint propagation is weaker than CSP but many issues are common to both.



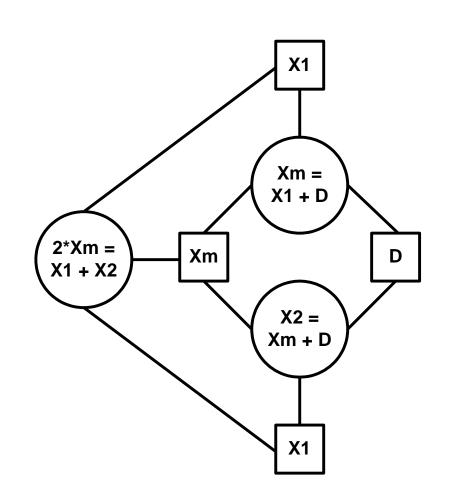
global and local consistency, relaxation

Example Propagation Network Problem



- Line segment mid-point:
 - $\mathbf{X}\mathbf{M} = \mathbf{X}\mathbf{1} + \mathbf{D}$
 - X2 = Xm + D
- X1 and Xm given then d and X2 easily calculated
- Local propagation fails when X1 and X2 given.
- Solutions:
 - "relaxation"
 - modify network

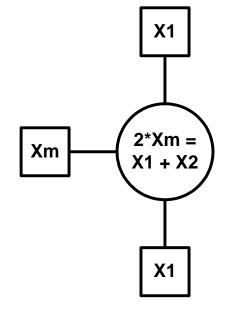
Solution 1: Augment Network



- Address problem by augmenting network.
- Add a new constraint
 - $2 \times Xm = X1 + X2$
- New constraint prevents local propagation from getting stuck
- New constraint mechanically derivable from existing constraints



Solution 2: Simplify Network



- Or address problem by simplifying network.
- Remove D and combine constraints involving D.
- Decision to remove D not suitable for mechanization.
- Simple solution in this case but other cases...



Conclusion

- SAR Inference Engine
 - A implements a single logical model
 - Avoids inconsistencies arising from multiple fragmented models
 - Focuses refinement and maturation efforts.
 - Is usable throughout lifecycle of radar
 - Design through deployed operation
 - Diverse uses and comprehensive scope enabled by
 - Constraint programming paradigm
 - Constraint propagation network

