





The IvP Helm and New Features of MOOS-IvP 4.2



Michael R. Benjamin

MIT Dept. of Mechanical Engineering Computer Science and Al Lab (CSAIL)

mikerb@mit.edu http://oceanai.mit.edu/mikerb



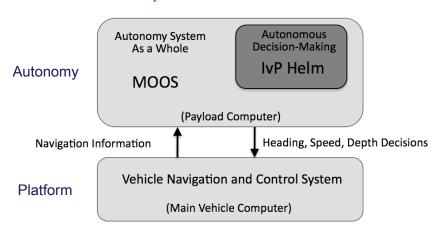


Payload UUV Autonomy



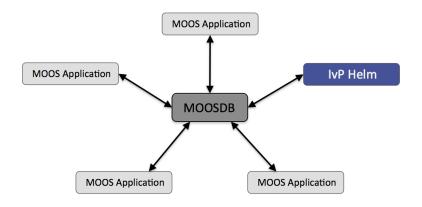
(3 Architecture Principles)

Principle #1 – Separation of Vehicle Autonomy from the Physical Platform

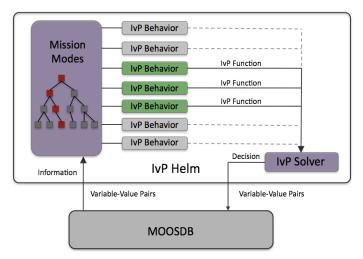


- Each architecture principle is designed to provide the end user with choices and the ability to procure capabilities on a component rather than system basis.
- Allows for incremental development across organizations.
- These new thrusts will result in tangible new software modules and capabilities.

Principle #2 – Separation of Autonomy System Components (MOOS Middleware)



Architecture Principle #3 – Separation of Autonomy into dedicated behaviors (IvP Helm)





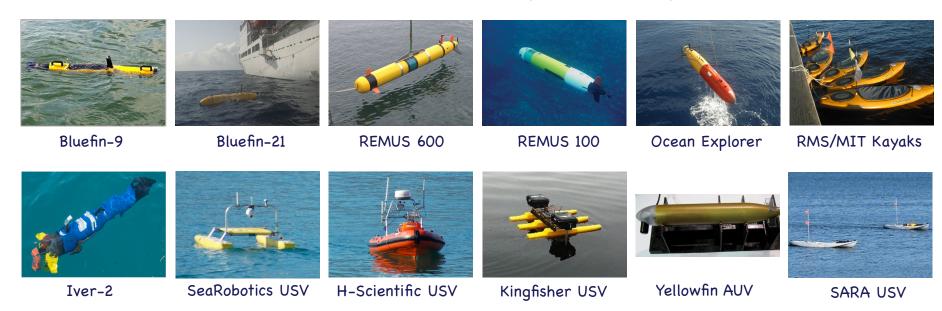


Overview of MOOS-IvP Payload Autonomy Software

Autonomy Software at MIT – Laboratory for Autonomous Marine Sensing Systems

- Open Source software project 70+ applications 130,000+ lines of code, 25-30 work years of research & development.
- MOOS-IvP autonomy software www.moos-ivp.org
- Goby underwater communications software <u>www.gobysoft.org</u>
- "An Overview of MOOS-IvP and a Brief Users Guide to the IvP Helm Autonomy Software" http://dspace.mit.edu/handle/1721.1/45569
- "Extending a MOOS-IvP Autonomy System and Users Guide to the IvPBuild Toolbox" http://dspace.mit.edu/handle/1721.1/46361
- "MOOS-IvP Autonomy Tools Users Manual" http://dspace.mit.edu/handle/1721.1/43708

Platforms that have field-demonstrated MOOS-IvP payload autonomy:



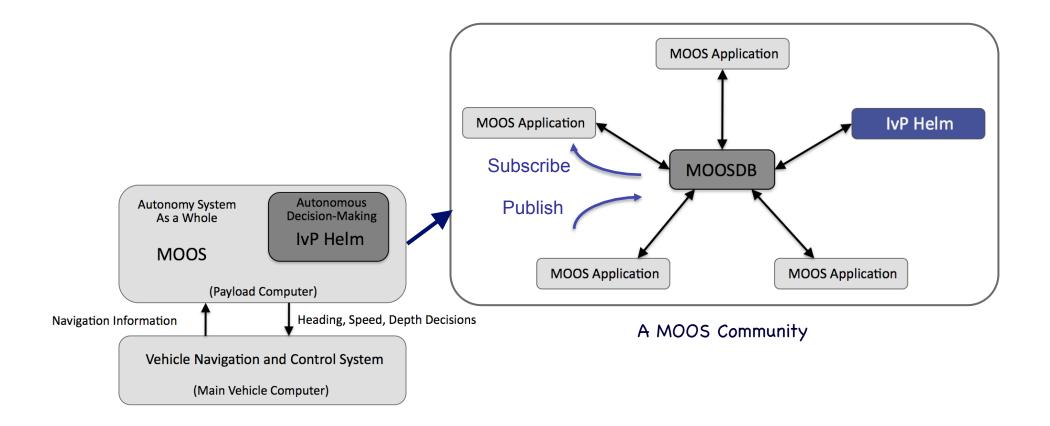


The Payload Autonomy Paradigm



Principle #2 - Separation of Autonomy System Components (MOOS)

- MOOS is middleware built on the publish-subscribe architecture.
- Each MOOS application is a separate process running on the vehicle computer.
- The interface of each process is defined by the messages it publishes and the messages it subscribes for.



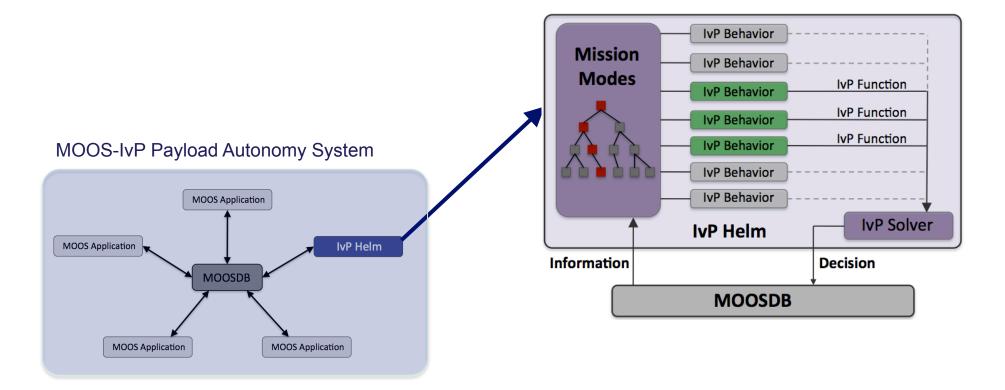


The Payload Autonomy Paradigm



Principle #3 - Separation of Autonomy into dedicated distinct behaviors.

- The IvP Helm is a decision-making engine based on the behavior-based architecture. It is a single MOOS application comprised of multiple specialized behaviors.
- Behaviors are turned on or off based on defined situations (states) and transitions. When multiple behaviors are active, coordination is by multi-objective optimization.
- Interval Programming (IvP) is the technique used for multi-objective optimization.







Outline

- Trends in autonomous marine vehicles
- The Payload Autonomy Paradigm and the MOOS-IvP project
- Multi-Objective Optimization with Interval Programming
- The IvP Helm
- MOOS-IvP 4.2 and Plans for Future Development

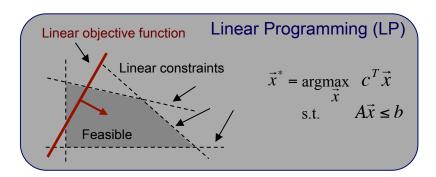


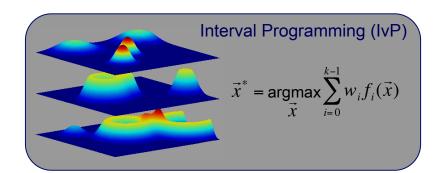
Interval Programming



Interval Programming is a mathematical programming model.

- A mathematical structure or syntax for representing an optmization problem.
- A toolbox of practical methods for casting different problems into the IvP syntax.
- A set of solution algorithms that exploit the syntactic structure.





The "IvP Helm" is a MOOS App implementing a behavior-based architecture for autonomous decision making, using Interval Programming to reconcile competing behaviors.

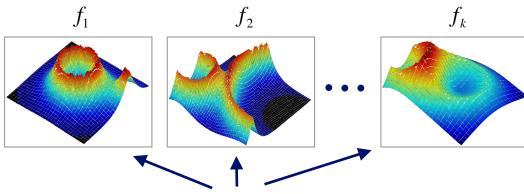


Interval Programming



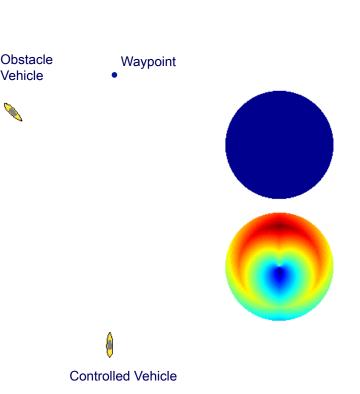
$$\vec{\chi}^* = \underset{\vec{\chi}}{\operatorname{argmax}} \sum_{i=0}^{k-1} w_i f_i(\vec{\chi})$$

- The solution, $\vec{\chi}^*$, is the single decision that maximizes the weighted sum of all utility functions.
- For example, the best combination of vehicle heading, speed, and depth.





- Each function is a utility function a mapping from possible decisions to their relative merit.
- Freedom from function form assumptions is key to solving general autonomy problems.
- Global optimality guaranteed.
- New problem generated and solution found on each iteration of the decision loop, typically 4x per second.

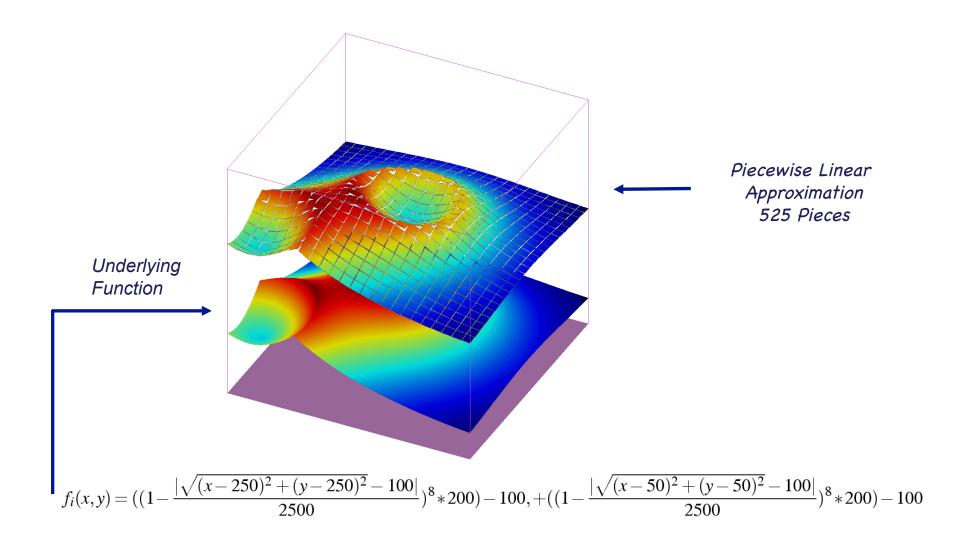




IvP Functions



An IvP Function is a piecewise linear approximation of an objective function, over a discrete decision space.

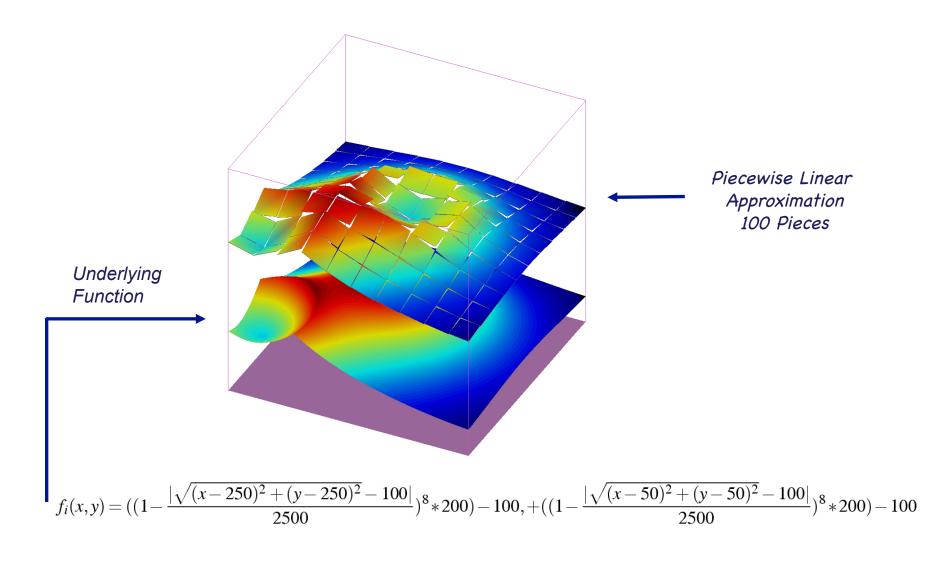




IvP Functions



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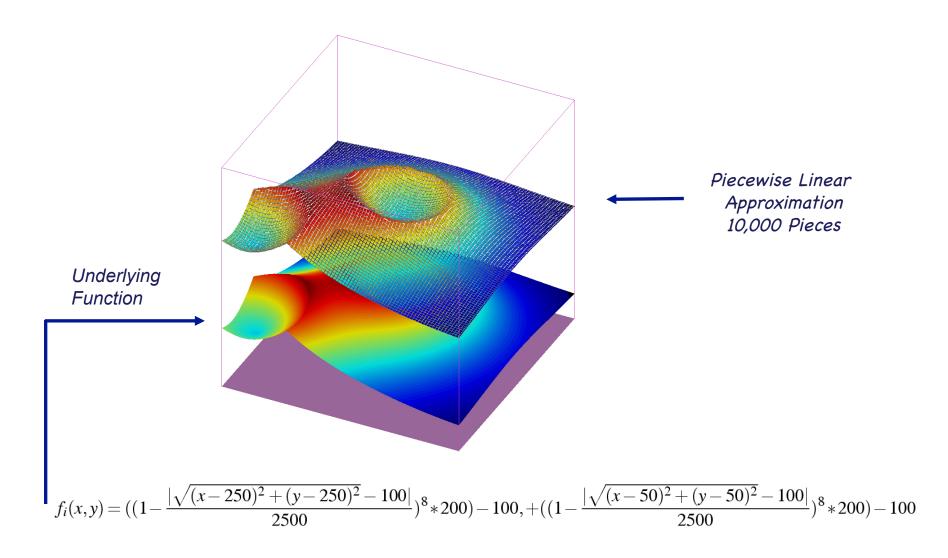




IvP Functions



An IvP Function is a piecewise linear approximation of an objective function, over a discrete decision space.





Piecewise Linear Functions in IvP

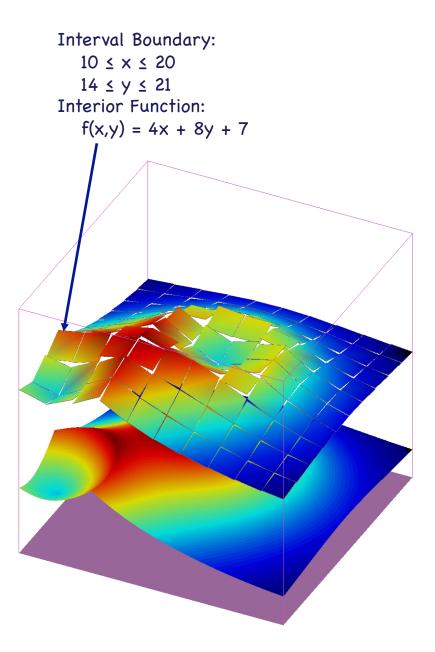


Piecewise linear (IvP) functions:

- Each point in the decision space belongs to exactly one piece.
- Each pieces has an interval boundary and a linear interior function.

Advantages:

- Any underlying function can be represented.
- Pieces need not be uniformly distributed.
- Extends to *n* dimensions.
- Syntax can be exploited by the solution algorithms.



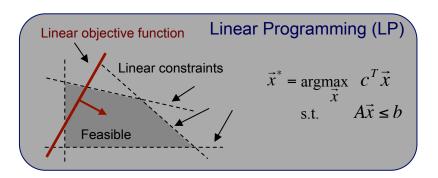


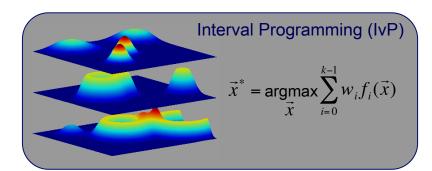
Interval Programming



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- A mathematical structure or syntax for representing an optmization problem.
- •A toolbox of practical methods for casting different problems into the IvP syntax.
 - A set of solution algorithms that exploit the syntactic structure.





The IvPBuild Toolbox:

- A C++ library of tools for building syntactically correct IvP functions.
- Typically invoked from code within an IvP Helm behavior implementation
- Downloadable from www.moos-ivp.org
- Documentation: http://dspace.mit.edu/handle/1721.1/46361



The IvPBuild Toolbox



Q: How are IvP Functions built?

A: The IvPBuild Toolbox

The IvPBuild Toolbox is a

- C++ Library,
- Distributed with the MOOS-IvP tree.
- A set of tools for building IvP functions from a user's underlying objective function.
- Meant to be invoked from within a behavior implementation from within the onRunState() function.
- The IvPBuild Toolbox contains two basic tools:
- The ZAIC tool for building 1D objective functions.
- The Reflector tool for building IvP Functions in N dimensions.



The Reflector Tool



(Pure Uniform)

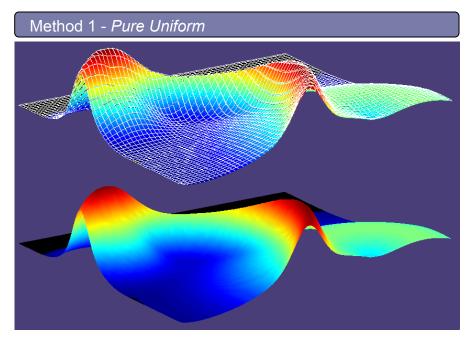
The Reflector Tool builds an IvP function from a given underlying function by sampling the underlying function.

Three variations discussed:

- → (1) Pure Uniform
 - (2) Uniform with Prioritized Augmentation
 - (3) Uniform with Focused Augmentation

Algorithm Design Criteria

- (1) Minimize error.
- (2) Minimize pieces generated.
- (3) Minimize generation time.
- \bigstar (4) Minimize complexity of use for the user.
- (5) Minimize restrictiveness of the tool.



Basic idea:

• Function is composed of uniform piecewise linearly defined pieces.

Pros:

- Simple to use.
- Requires no insight into underlying function.
- Can explore time, size, accuracy tradeoff space.

Cons:

- Treats all areas of the underlying function equally.
- Does not capitalize on insight into underlying function.



The Reflector Tool



(Uniform with Prioritized Augmentation)

The Reflector Tool builds an IvP function from a given underlying function by sampling the underlying function.

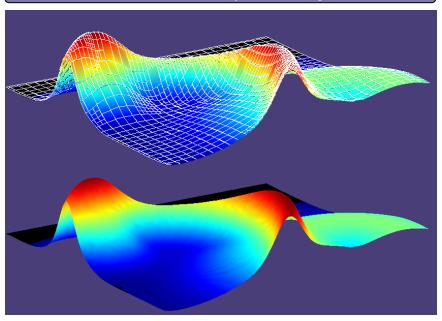
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Algorithm Design Criteria

- ★ (1) Minimize error.
- ★ (2) Minimize pieces generated.
 - (3) Minimize generation time.
- \bigstar (4) Minimize complexity of use for the user.
 - (5) Minimize restrictiveness of the tool.

Method 2 - Uniform with Priority-Based Augmentation



Basic idea:

Start with a uniform function and further refine the pieces that have the worst error (prioritized during first linear regression phase).

Pros:

- Simple to use. No insight into underlying function required
- Can explore time, size, accuracy tradeoff space.

Cons:

- Does not always catch the pieces with worst error.
- Does not capitalize on insight into underlying function.



The Reflector Tool



(Uniform with Focused Augmentation)

The Reflector Tool builds an IvP function from a given underlying function by sampling the underlying function.

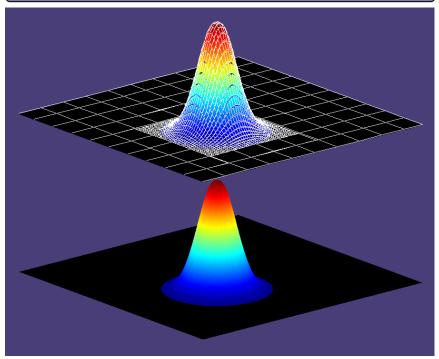
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- (1) Pure Uniform
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Algorithm Design Criteria

- ★ (1) Minimize error.
- 🜟 (2) Minimize pieces generated.
- \bigstar (3) Minimize generation time.
 - (4) Minimize complexity of use for the user.
 - (5) Minimize restrictiveness of the tool.

Method 3 - Uniform with Focused Augmentation



Basic idea:

Start with a uniform function and further refine the pieces in areas thought to need more pieces for error reduction

Pros:

- Simple to use. Capitalizes on insight of underlying function.
- Can explore time, size, accuracy tradeoff space.

Cons:

- Not all functions have area suitable for focused refinement.
- Requires insight into underlying function.







The ZAIC Tool builds an IvP function from a given underlying function by sampling the underlying function.

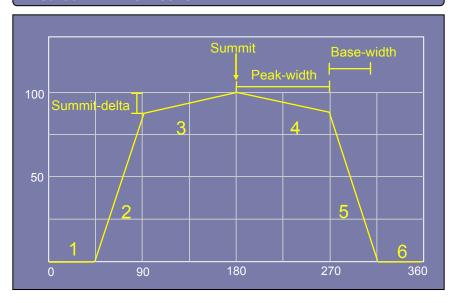
Three variations:

- (1) ZAIC_PEAKS
- (2) ZAIC_LEQ
- → (3) ZAIC_HEQ

Algorithm Design Criteria

- ★ (1) Minimize error.
- 🜟 (2) Minimize pieces generated.
- ★ (3) Minimize generation time.
- \bigstar (4) Minimize complexity of use for the user.
- \bigstar (5) Minimize restrictiveness of the tool.

Method 4 - ZAIC Peaks



Basic idea:

1D Functions with one or more peaks. Identify the peak properties and the IvP function is generated.

Pros:

Simple to use. Very few pieces. As many peaks as desired.

Cons:

Only suitable for 1D objective functions.

IvP Function:

- $(1) 0 \le x \le 45 y = 0$
- (2) $46 \le x \le 90$ y = 1.89x 85
- (3) $91 \le x \le 180$ y = 0.17x + 70
- (4) $181 \le x \le 270$ y = -0.17x + 130(5) $271 \le x \le 315$ y = -1.89x + 595
- (6) $316 \le x \le 359$ y = 0

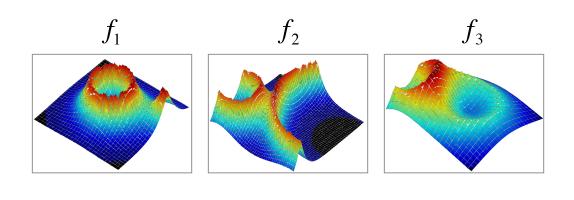


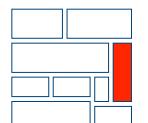
Interval Programming Solution Algorithms (Overview)

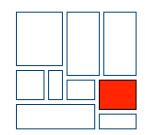


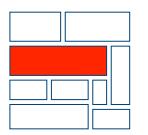
An IvP problem consists of a set of k functions, each with a priority weighting. The solution is given by:

$$\vec{\chi}^* = \underset{\vec{\chi}}{\operatorname{argmax}} \sum_{i=0}^{k-1} w_i f_i(\vec{\chi})$$







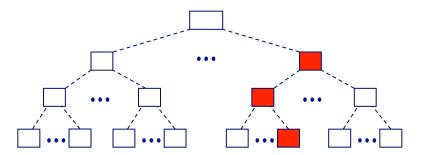


The Search Tree:

- 1 level for each function
- n^k leaf nodes (n pieces per function).

The Solution algorithm:

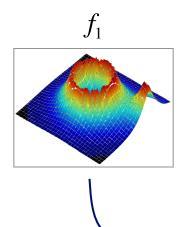
- Branch and bound
- Pruning based on intersection lookahead.

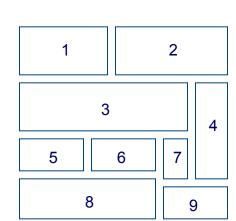


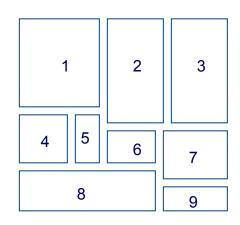


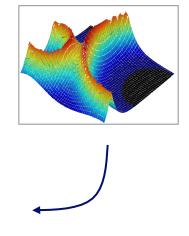












How global optimality is guaranteed:

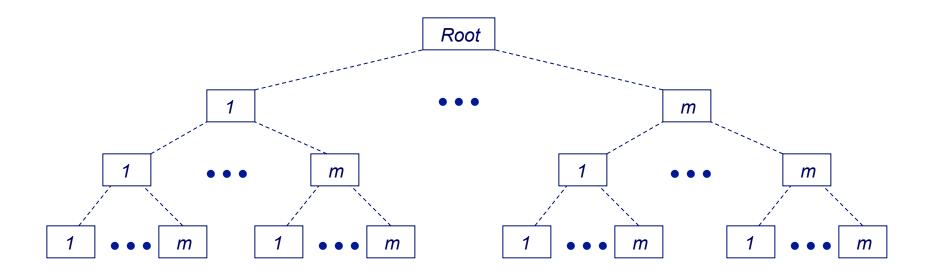
- Each objective function is defined over the same decision space (x, y), (heading, speed) etc.
- The solution space is the set of all possible combinations of pieces from each function.

- Each point in the decision space belongs to exactly one piece from each piecewise function.
- Therefore considering all possible combinations ensures all possible decisions are considered.



IvP Solution Algorithms (The Search Tree)





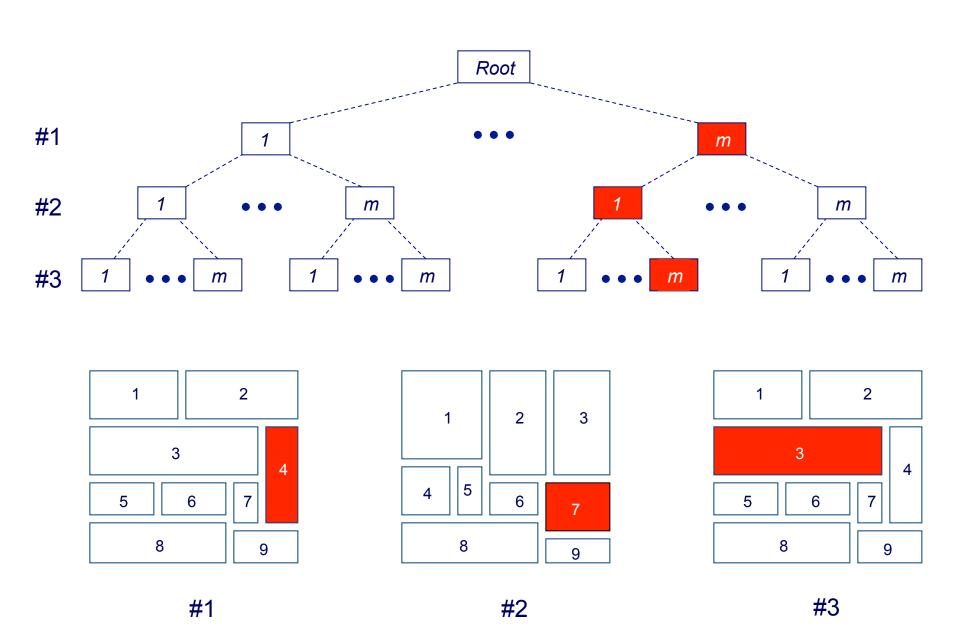
- Each level in the tree corresponds to one objective function.
- Each leaf node is a combination of one piece from each objective function.
- Function weights applied a priori to the pieces weights no longer relevant.

k functions, m pieces: m^k leaf nodes!





Closer Look at a Leaf Node



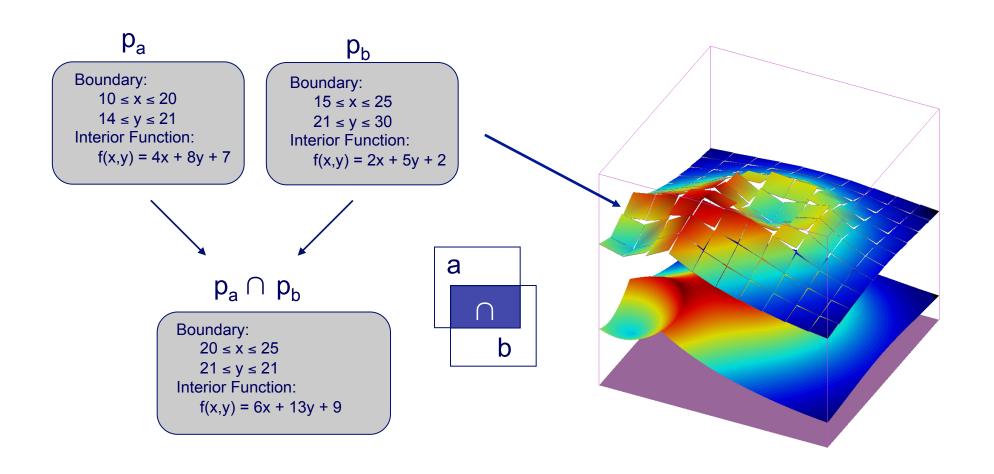




(Piece Intersection)

Two pieces intersect if their boundaries overlap.

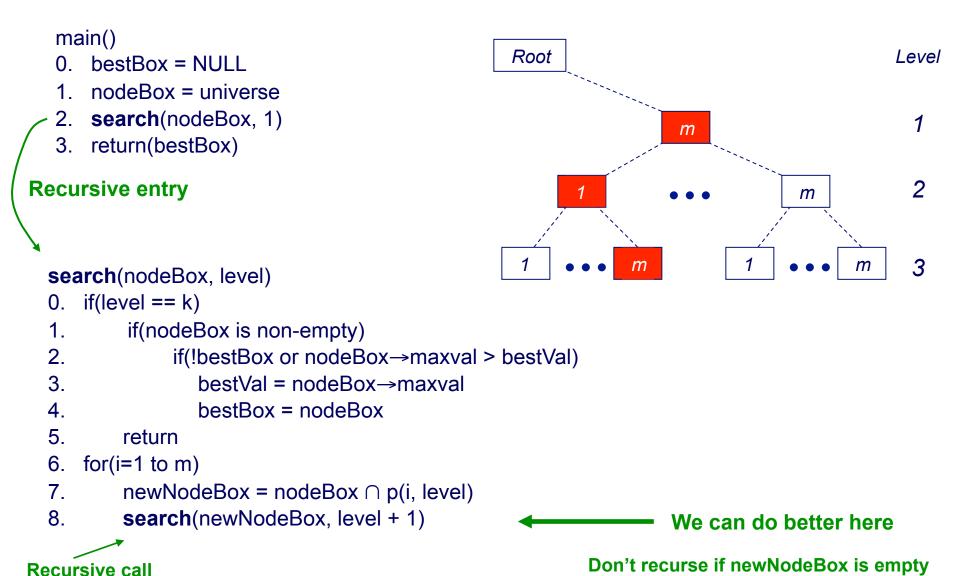
The result is a new piece with common boundary and the sum of their two linear interior functions.







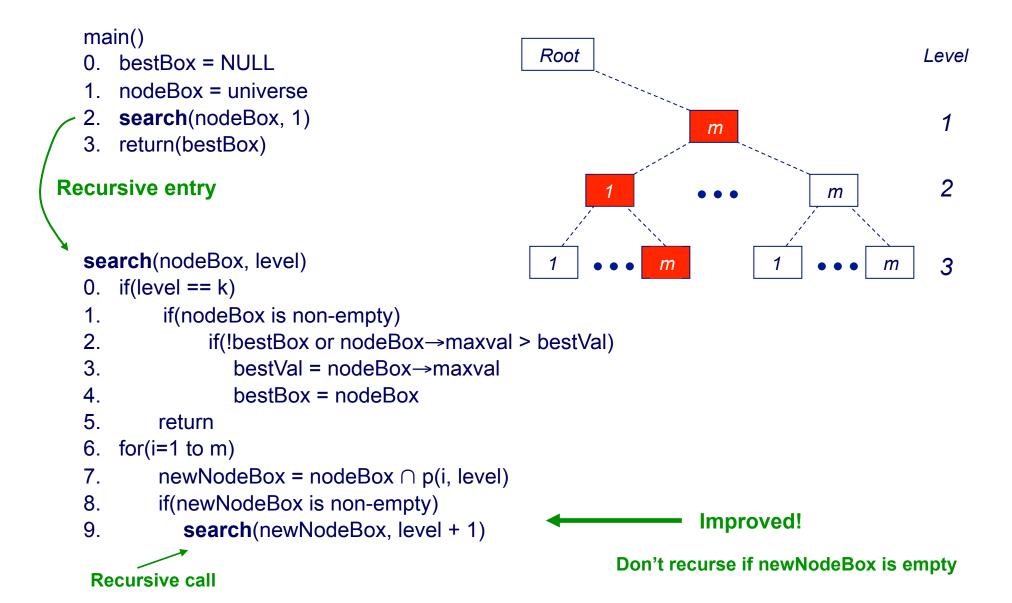
(Full Search Tree Expansion)







(Simple Pruning)

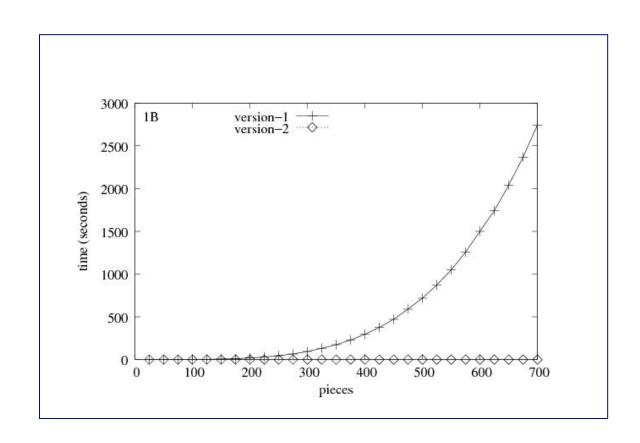








(Full Expansion vs. Simple Pruning)

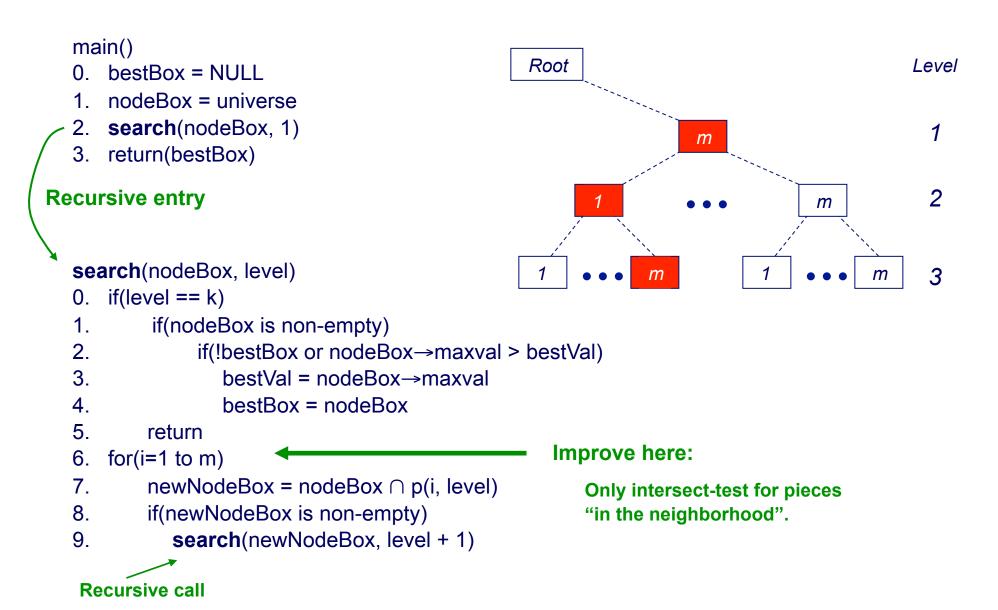


Each problem contains 3 objective functions 3 dimensions





(Simple Pruning)



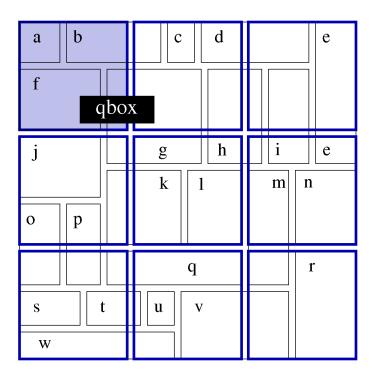




(Efficient Intersection Detection)

Only intersect-test for pieces "in the neighborhood".

- A grid is associated with each piecewise objective function.
- A list is associated with each grid element containing pieces that intersect that element.
- The grid is populated when the function is constructed before the solution phase begins.
- The grid configuration can be different for each objective function.

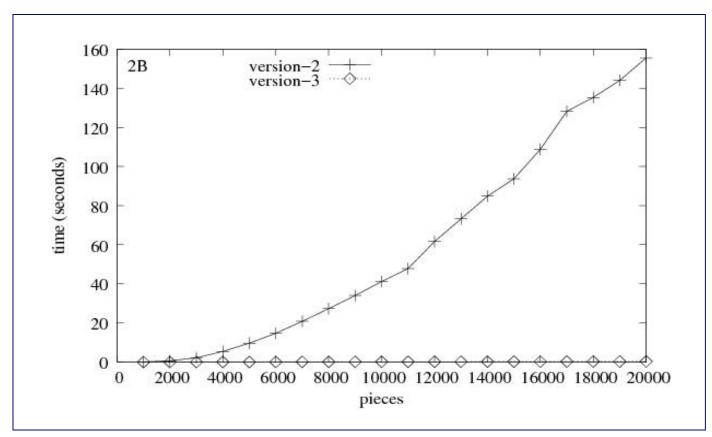


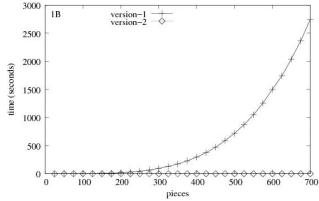
 $Grid(level) \rightarrow getBoxes(qbox) = \{a, b, c, d, f, g, h\}$





(Exhaustive vs. Grid-based Intersection Testing)





Each problem contains 3 objective functions 3 dimensions





(Internal Node Upper Bounds)

Get the upper bound for pieces "in the neighborhood".

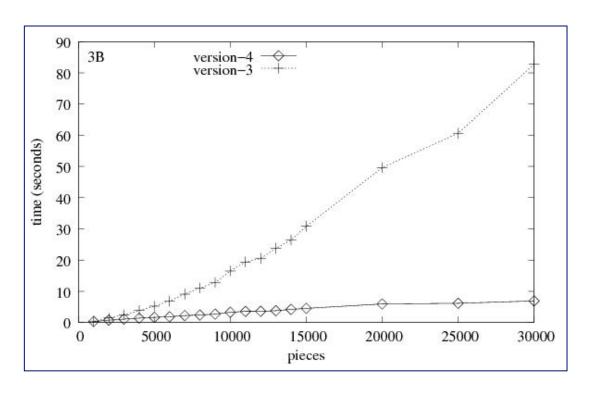
- Each piece has a known maximum value (of its linear interior function).
- Each grid element stores the max value of all pieces added to that element
- The grid is populated when the function is constructed before the solution phase begins.
- The grid configuration can be different for each objective function.

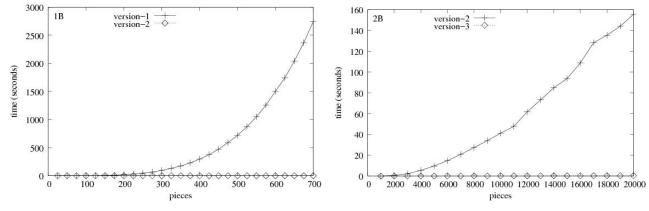
22	1		6 2	1			2
11 qbo			17 4		12		
14							
		8	27		2 12		2
10	11						
			15		27		
5 16		5	32				
3							





(No-Upper-Bound vs. Upper-bound)





Each problem contains 3 objective functions 3 dimensions



Outline

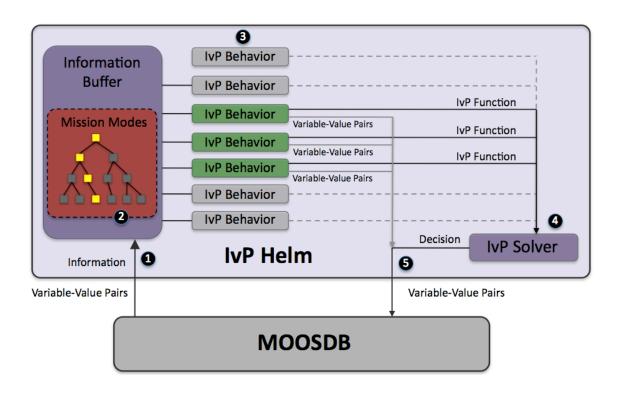


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Interval Programming and the IvP Helm



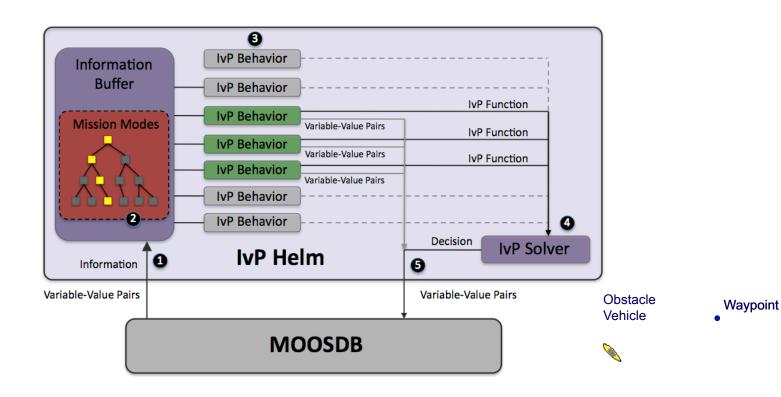


- 1 Mail is read in the MOOS OnNewMail() function and applied to a local buffer.
- 2 The helm mode is determined, and set of running behaviors determined.
- 3 Behaviors do their thing posting MOOS variables and an IvP function.
- 4 Competing behaviors are resolved with the IvP solver.
- 5 The Helm decision and any behavior postings are published to the MOOSDB.

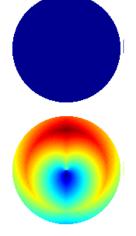


Interval Programming and the IvP Helm





- 1 Mail is read.
- 2 Helm mode is determined.
- 3 Behaviors generate their output.
- 4 Competing behaviors are resolved.
- 5 The Helm posts its results

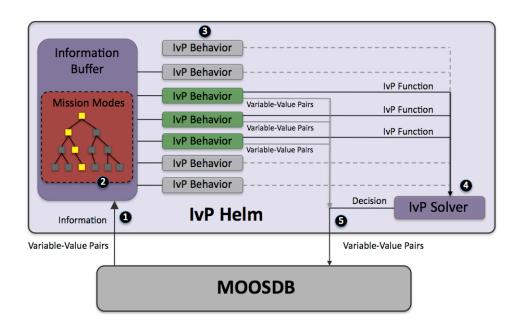






IvP Helm Configuration





Helm configuration file structure:

Variable Initializations

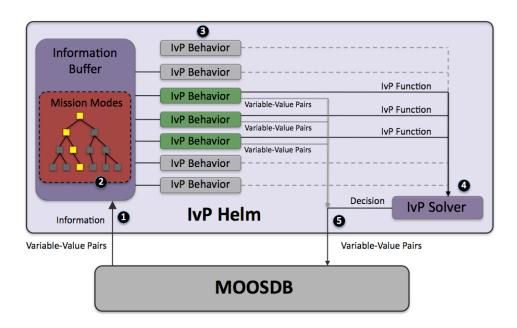
Hierarchical Mode Declarations

Behavior Configurations



IvP Helm Configuration





Variable Initializations

Hierarchical Mode Declarations

Behavior Configurations





Mission Synopsis:

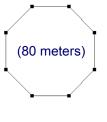
Upon receiving a deploy command, transit to and loiter at region A for a fixed duration and then to region B. Periodically switch between regions until recalled home.

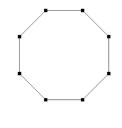
```
Behavior = BHV_Loiter
{
   parameter = value
   . . .
   parameter = value
}
```

```
Behavior = BHV_Return
{
   parameter = value
   . . .
   parameter = value
}
```

Launch and return position







REGION A

REGION B





Mission Synopsis:

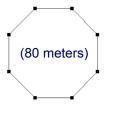
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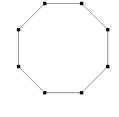
```
Behavior = BHV_Return
{
  name = return
  condition = (DEPLOY=true) and (RETURN=true)

  speed = 1.8
  radius = 4.0
  point = 80,40
}
```

Launch and return position







REGION A

REGION B





Mission Synopsis:

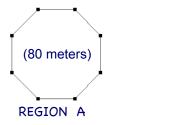
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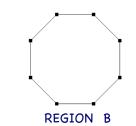
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{
  name = return
  condition = (DEPLOY=true) and (RETURN=true)

  speed = 1.8
  radius = 4.0
   point = 80,40
}
```

Launch and return position







```
Initialize DEPLOY = false
Initialize RETURN = false
Initialize REGION = A
```

2

Variable Initializations

Hierarchical Mode Declarations

Behavior Configurations

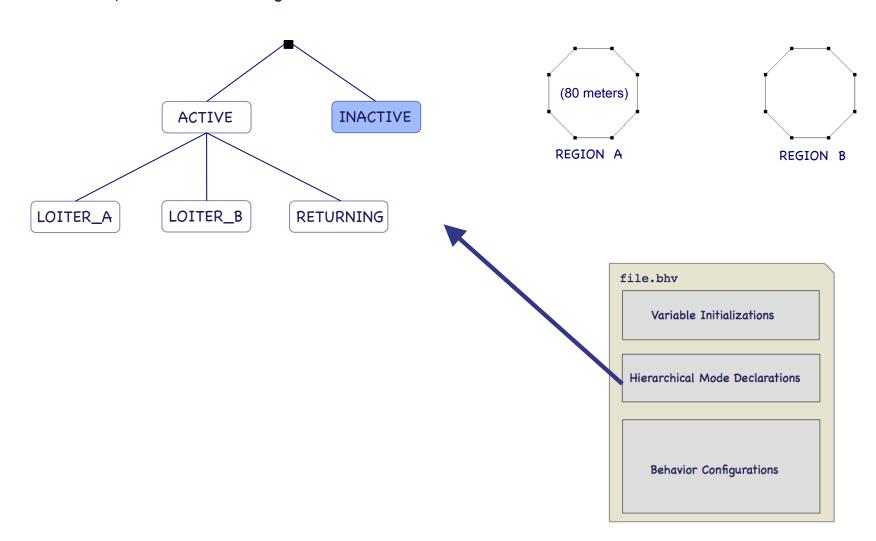




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Upon receiving a deploy command, transit to and loiter at region A for a fixed duration and then to region B. Periodically switch between regions until recalled home.

Launch and return position



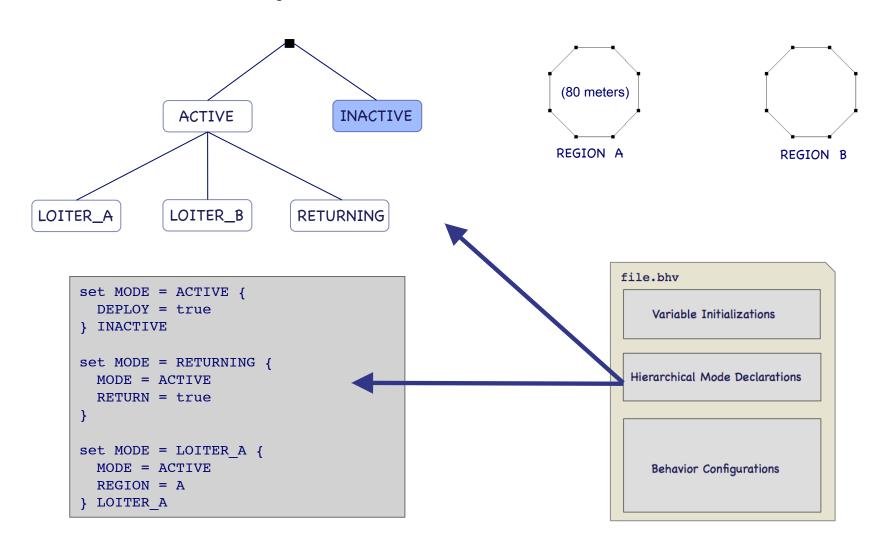




Mission Synopsis:

Upon receiving a deploy command, transit to and loiter at region A for a fixed duration and then to region B. Periodically switch between regions until recalled home.

Launch and return position

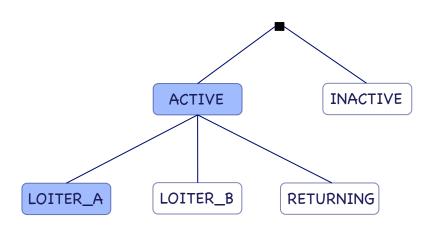






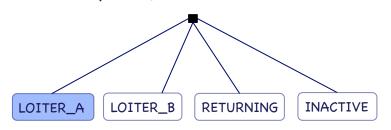
Mission Synopsis:

Upon receiving a deploy command, transit to and loiter at region A for a fixed duration and then to region B. Periodically switch between regions until recalled home.



Question:

Why define the "Active" mode? Why not just have:



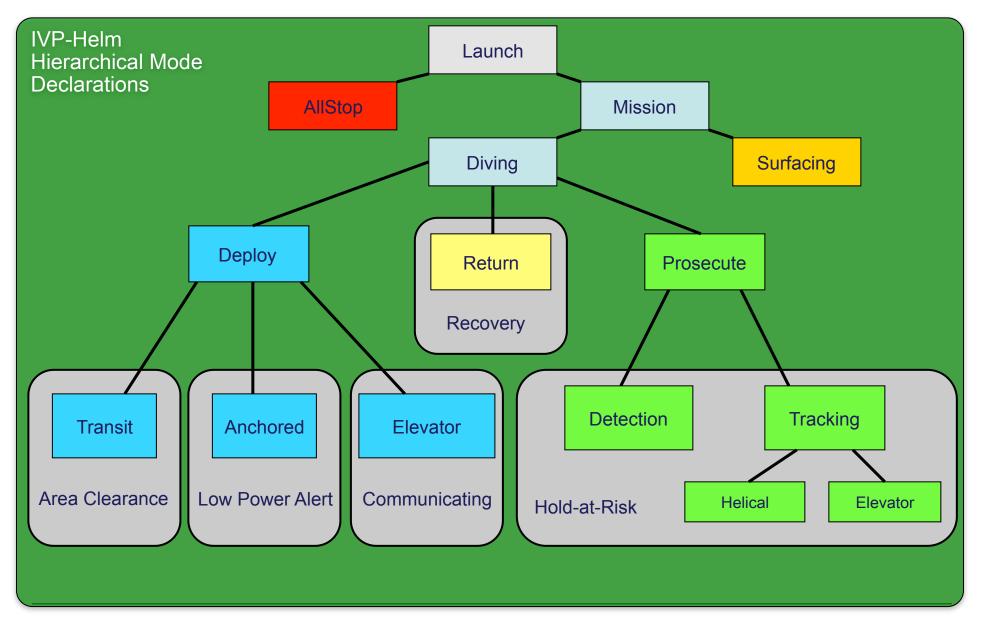
```
set MODE = ACTIVE {
    DEPLOY = true
} INACTIVE

set MODE = RETURNING {
    MODE = ACTIVE
    RETURN = true
}

set MODE = LOITER_A {
    MODE = ACTIVE
    REGION = A
} LOITER_A
```



MIT Prototype Autonomy Modes







Dynamic Modifications to the Helm

Q: How is the helm modified after launch?

A: By receipt of incoming MOOS mail.

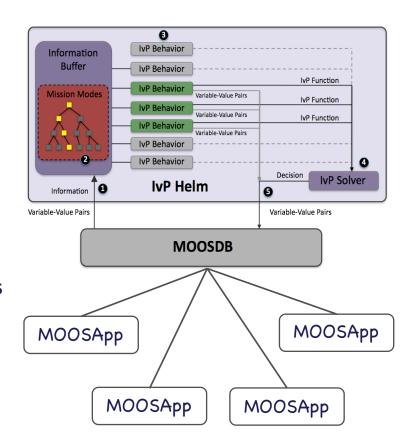
- The helm's mission mode may change
- Behavior parameters may change

Q: Why would the helm be modified?

- New internal plans generated
- Conclusions from sensor processing modules
- External comms from other vehicles
- External comss from field-control

Q: Why is this important?

A: It determines how the helm may interface with an off-board planner, field-control system, scheduler, and other vehicles.







Dynamic Modifications to the Helm

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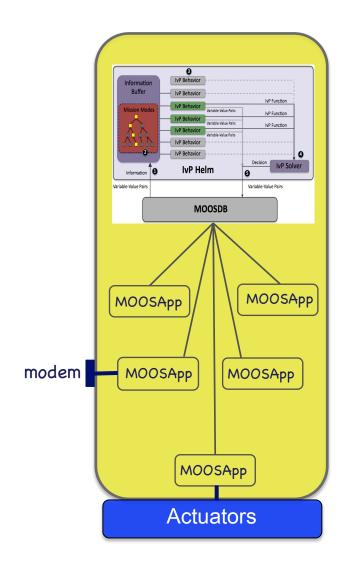
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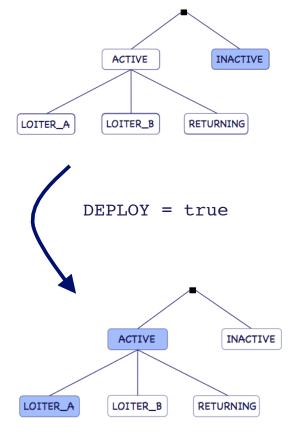






Dynamic Modifications to the Helm (1) Mode Modifications

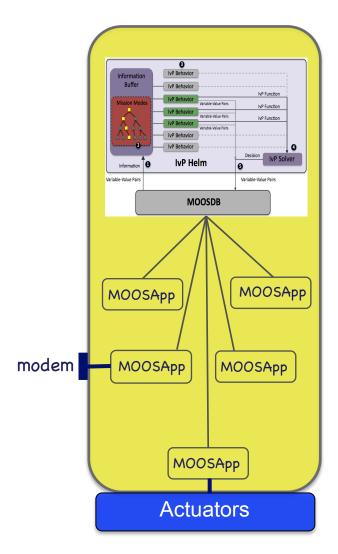
- The Helm Mode is determined by conditions
- The conditions are defined over MOOS variables:



```
set MODE = ACTIVE {
   DEPLOY = true
} INACTIVE

set MODE = RETURNING {
   MODE = ACTIVE
   RETURN = true
}

set MODE = LOITER_A {
   MODE = ACTIVE
   REGION = A
} LOITER_A
```





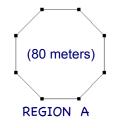
Dynamic Modifications to the Helm (2) Behavior Modifications

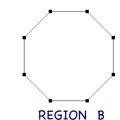


- Behavior configuration parameters may altered after launch.
- Each behavior may specify an UPDATES variable.

Launch and return position







```
LOITER_UPDATE = polygon=format=radial, x=160, y=-75, radius=40, pts=8
```







What is Dynamic Behavior Spawning?

- Behaviors may be defined as templates with instances spawned upon receipt of an externally generated, user-defined event.
- Behavior authors may implement behaviors to die under certain conditions, and post MOOS messages immediately prior to dying.

Motivation:

- For certain behaviors, e.g., collision avoidance, contact tracking, multiple instances of the behavior are required, one for each contact.
- It's virtually impossible to know the amount or type of contacts encountered prior to the start of the mission.





Configuring Behaviors with Dynamic Behavior Spawning

Non-Templated Behavior:

```
Behavior = BHV AvoidCollision
            = avd collision
 name
             = 200
 pwt
 condition = AVOID=true
             = CONTACT INFO
 updates
                contact = macrura
 active outer distance = 50
 active inner distance = 20
    completed distance = 75
    collision distance = 8
    all clear distance = 25
           active grade = linear
      on no contact ok = true
            extrapolate = true
                  decay = 30,60
```

Templated Behavior:

```
Behavior = BHV AvoidCollision
             = avd collision
  name
             = 200
  pwt
  condition = AVOID=true
            = CONTACT_INFO
 updates
             = CONTACT RESOLVED = $[CONTACT]
  endflag
  templating = spawn
                contact = to-be-set
  active outer distance = 50
  active inner distance = 20
     completed distance = 75
     collision distance = 8
     all clear distance = 25
           active grade = linear
       on no contact ok = true
            extrapolate = true
                  decay = 30,60
}
```

```
MOOS Post CONTACT_INFO = "name=avd_macrura # contact=macrura"

MOOS Post CONTACT_INFO = "name=avd_henry # contact=henry"
```





The Berta Example Mission with Dynamic Behavior Spawning

The Berta example mission:

- In moos-ivp/trunk/missions/m2_berta
- Two vehicles loitering and repeatedly swapping loiter locations
- Each time the vehicles get close, a collision avoidance behavior is spawned.
- After the range opens sufficiently, the collision avoidance behavior dies.







Monitoring Life Events

- A "Life Event" is the spawning or death of a behavior.
- Life Events may be monitored in a special mode of the uHelmScope MOOS utility:

\$ uHelmScope --life henry.moos

000			Terr	minal — uHelmScope — 86×	21 — 第1	
8	tcsh		⊗ uHe	lmScope		
**	*****	*****	******	******	****	
*		Summar	y of Behavi	or Life Events	*	
**	*****	*****	*****	*****	****	
Time	Iter	Event	Behavior	Behavior Type	Spawning Seed	
280.50	833	death	avd_gilda	BHV_AvoidCollision		
389.77	1268	spawn	avd_gilda	BHV_AvoidCollision	name=avd_gilda#contact=gilda	
444.52	1415	death		BHV_AvoidCollision		
557.64	1867	spawn		BHV_AvoidCollision	name=avd_gilda#contact=gilda	
617.70	2019	death		BHV_AvoidCollision		
701.60	2355	spawn			name=avd_gilda#contact=gilda	
758.22	2511		avd_gilda	_		
809.56	2717	spawn			name=avd_gilda#contact=gilda	
866.61 971.80	2863 3273		avd_gilda	_	nome-oud wilde#gontogt-wilde	
1031.60	3410	spawn death		BHV_AvoidCollision	name=avd_gilda#contact=gilda	
1031.00	3410	death	avu_giida	PHA_WANTECOTTIBION		
Hit 'r'	to res	ume out	puts, or SP	ACEBAR for a single	update	Ā
			F = 35, OF BI	zoz w bangzo		*
						11.



Analyzing Life Events



- A "Life Event" is the spawning or death of a behavior.
- Life Events may be monitored in a special mode of the uHelmScope MOOS utility.
- The Life Event may also be examined post-runtime from the MOOS log files:

\$ aloghelm --life henry_logfile.alog

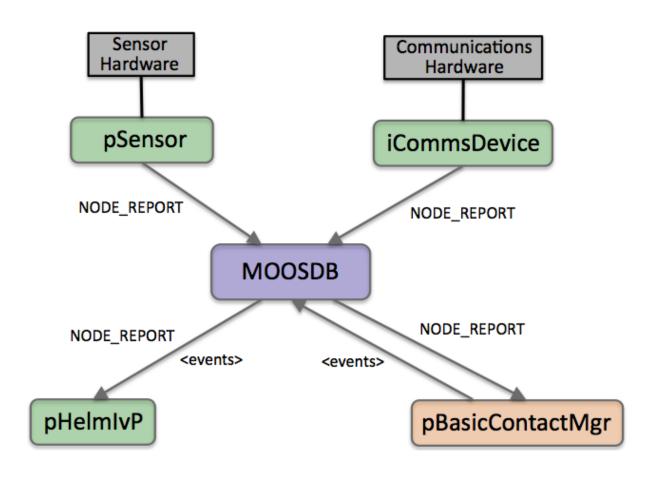
8	sh			PID	0	tcsh				
** * **	*****		************* y of Behavior *******			*** * **				
Time	Iter	Event	Behavior	Behavior	Туре	Spawning S	eed			
0.00	1	spawn	loiter	BHV_Loit		helm start	up			
0.00	1	spawn	waypt_return	BHV_Wayp	oint	helm start	up			
0.00	1	spawn	station-keep	BHV_Stat	ionKeep	helm start	up			
94.57	247	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
152.94	398	death	avd_gilda	BHV_Avoi	dCollision					
222.32	677	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
280.50	833	death	avd_gilda	BHV_Avoi	dCollision					
389.77	1268	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
444.52	1415	death	avd_gilda	BHV_Avoi	dCollision					
557.64	1867	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
617.70	2019	death	avd_gilda	BHV_Avoi	dCollision					
701.60	2355	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
758.22	2511	death	avd_gilda	BHV_Avoi	dCollision					
809.56	2717	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
866.61	2863	death	avd_gilda	BHV_Avoi	dCollision					
971.80	3273	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
1031.60	3410	death	avd_gilda	BHV_Avoi	dCollision					
1164.46	3927	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
1226.22	4061	death	avd_gilda	BHV_Avoi	dCollision					
1341.33	4503	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd_g	ilda#contact=gilda			
1400.98	4638	death	avd_gilda	BHV_Avoi	dCollision					
1520.36	5108	spawn	avd_gilda	BHV_Avoi	dCollision	name=avd g	ilda#contact=gilda	1		
leonardo	:missi	ons/m2_	berta(trunk)%					<u> </u>		





Contact Management and Behavior Spawning

- A MOOS application pBasicContactMgr.
- It receives NODE_REPORT messages from other MOOS applications





Non-Traditional Aspects of Behavior-Based Control in the IvP Helm



- · Behaviors have state.
- Behaviors influence each other between iterations.
- Behaviors accept externally generated plans.
- There may be several instances of the same behavior.
- Behaviors may spawn and die dynamically based on events or commands.
- Behaviors may run in a configurable sequence.
- Behaviors rate actions over a coupled decision space (multi-objective optimization)



This is not Rodney Brooks' Behavior Based Control

However - the power of independent, incremental development has been retained, enhanced by the power of Open Source development, and wide, diverse collaborations.





(General Characteristics)

Purpose: Traverse a given set of waypoints, gracefully handling missed vertices.

Parameters:

points: A set of points in the X-Y plane

capture_radius: Distance from a point, within which arrival is declared.

speed: Desired speed of traversal.

non-monotonic_radius: Distance from a point, within which an increase in distance is treated as an arrival.

track_lead: For track-line following - distance to perpendicular intersection point.

order: Order of point traversal.

repeat: Number of times points are traversed.

Example:

points: (0,-80), (45,-45), (160, -120)

capture_radius: 4

speed: 2.5

Start position



80 meters:









(Non-monotonic Radius)

Purpose: Traverse a given set of waypoints, gracefully handling missed vertices.

Parameters:

points: A set of points in the X-Y plane

capture radius: Distance from a point, within which arrival is declared.

> Desired speed of traversal. speed:

Distance from a point, within which an increase in distance is treated as an arrival. non-monotonic_radius:

track_lead: For track-line following - distance to perpendicular intersection point.

order: Order of point traversal.

repeat: Number of times points are traversed.

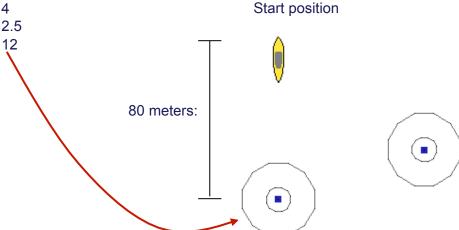
Example:

(0,-80), (45,-45), (160,-120)points:

capture radius:

2.5 speed:

non-monotonic radius









(Track-line Following)

Purpose: Traverse a given set of waypoints, gracefully handling missed vertices.

Parameters:

points: A set of points in the X-Y plane

capture_radius: Distance from a point, within which arrival is declared.

speed: Desired speed of traversal.

non-monotonic_radius: Distance from a point, within which an increase in distance is treated as an arrival.

track_lead: For track-line following - distance to perpendicular intersection point.

order: Order of point traversal.

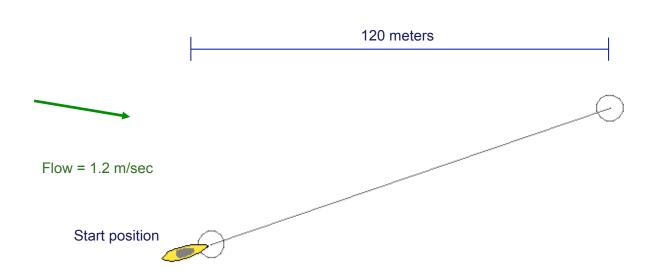
repeat: Number of times points are traversed.

Example:

points: (0,-45), (120,0)

capture_radius: 4

speed: 2.0 track lead: 0







(Track-line Following)

Purpose: Traverse a given set of waypoints, gracefully handling missed vertices.

Parameters:

points: A set of points in the X-Y plane

capture_radius: Distance from a point, within which arrival is declared.

speed: Desired speed of traversal.

non-monotonic_radius: Distance from a point, within which an increase in distance is treated as an arrival.

track_lead: For track-line following - distance to perpendicular intersection point.

order: Order of point traversal.

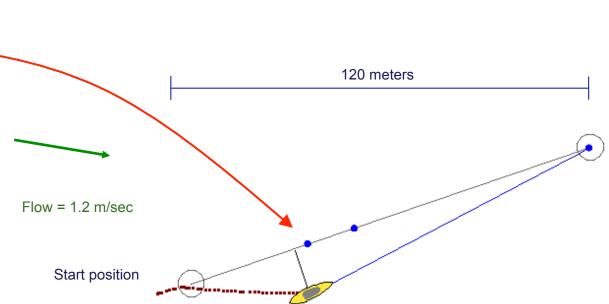
repeat: Number of times points are traversed.

Example:

points: (0,-45), (120,0)

capture_radius: 4

speed: 2.0 track lead: 2







The Obstacle Avoidance Behavior

Purpose: Avoid a set of given obstacles each represented by a *convex* polygon.

Parameters:

polygon: A set of points in the X-Y plane, comprising a convex polygon.

polygon: Other obstacles.

allowable_ttc: Time To Collision Allowed before a candidate maneuver is penalized.

activation_dist: Distance to a polygon beyond which the behavior is inactive.

buffer_dist: Distance to polygon treated as a collision.

Example: polygon: (10,10), (20,20), (30,30), (40,40)

polygon: (60,60), (70,70), (80,80), (90,90)

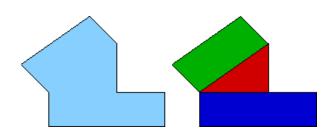
allowable_ttc: 25 activation_dist: 60 buffer_dist: 8

Q: Why convex polygons?

A: Most operations are simplified.

- Point containment test,
- Distance to polygon,
- Line segment intersection, etc.

A: A non-convex polygon can be represented by a set of convex polygons.



Start







Waypoint





The Obstacle Avoidance Behavior

Purpose: Avoid a set of given obstacles each represented by a *convex* polygon.

Parameters:

polygon: A set of points in the X-Y plane, comprising a convex polygon.

allowable ttc: Time To Collision Allowed before a candidate maneuver is penalized.

activation dist: Distance to a polygon beyond which the behavior is inactive.

buffer_dist: Distance to polygon treated as a collision.

Example: polygon: (10,10), (20,20), (30,30), (40,40)

polygon: (60,60), (70,70), (80,80), (90,90)

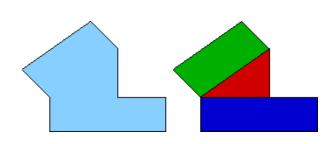
allowable_ttc: activation dist: 60 buffer_dist: 8

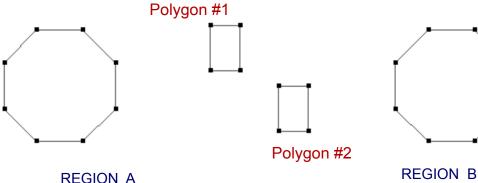
Q: Why convex polygons?

A: Most operations are simplified.

- Point containment test.
- Distance to polygon,
- Line segment intersection, etc.

A: A non-convex polygon can be represented by a set of convex polygons.









The Loiter Behavior

Purpose: Repeatedly traverse a given set of waypoints, gracefully handling missed vertices.

Automatically calculate trajectory re-entry when required.

Parameters:

polygon: A set of points in the X-Y plane, comprising a convex polygon.

capture radius: Distance from a point, within which arrival is declared.

> Desired speed of traversal. speed:

non-monotonic radius: Distance from a point, within which an increase in distance is treated as an arrival.

acquire_distance: Distance from the polygon, outside of which the behavior is in "acquire mode".

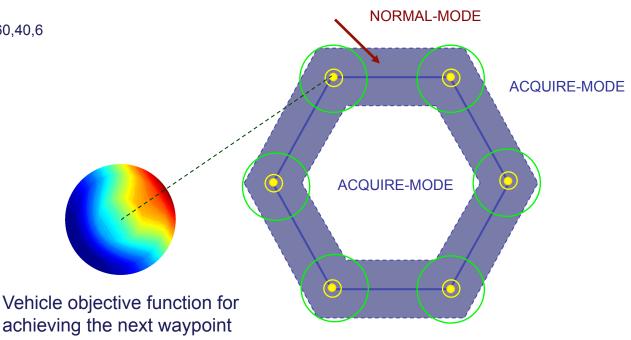
clockwise: True if traversing clockwise.

Example:

radial:50,60,40,6 polygon:

capture_radius: 10 non-monotonic_radius: 15 acquire_distance: 15

clockwise: true









(Acquire Vertex Policy - External Case)

Purpose: Repeatedly traverse a given set of waypoints, gracefully handling missed vertices.

Automatically calculate trajectory re-entry when required.

Parameters:

polygon: A set of points in the X-Y plane, comprising a convex polygon.

capture_radius: Distance from a point, within which arrival is declared.

speed: Desired speed of traversal.

non-monotonic_radius: Distance from a point, within which an increase in distance is treated as an arrival.

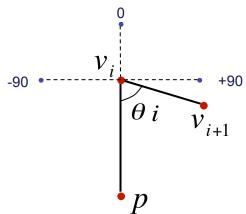
acquire_distance: Distance from the polygon, outside of which the behavior is in "acquire mode".

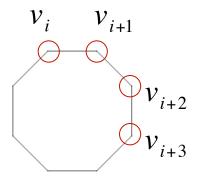
clockwise: True if traversing clockwise.

p

Acquire Vertex Policy (External Case):

acquire_vertex = vi where $i = \operatorname{argmin}(\theta_i)$ v_i is viewable from p





200 meters







(Acquire Vertex Policy - Internal Case)

Purpose: Repeatedly traverse a given set of waypoints, gracefully handling missed vertices.

Automatically calculate trajectory re-entry when required.

Parameters:

polygon: A set of points in the X-Y plane, comprising a convex polygon.

capture_radius: Distance from a point, within which arrival is declared.

speed: Desired speed of traversal.

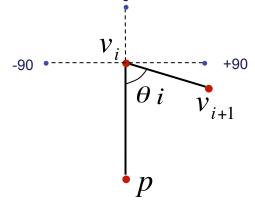
non-monotonic_radius: Distance from a point, within which an increase in distance is treated as an arrival.

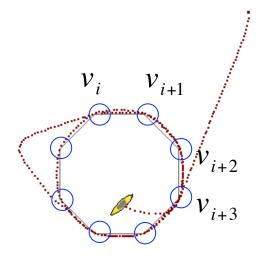
acquire_distance: Distance from the polygon, outside of which the behavior is in "acquire mode".

clockwise: True if traversing clockwise.

Acquire Vertex Policy (Internal Case):

acquire_vertex = v_i where $i = \operatorname{argmin}(\theta_i + k_i)$ $k_i = -c$ if v_i is current 0 otherwise





200 meters





The Turn-Limit Behavior

Purpose: To limit the rate of vehicle turn to protect a towed sensor.

Parameters:

memory_time: A set of points in the X-Y plane, comprising a convex polygon.

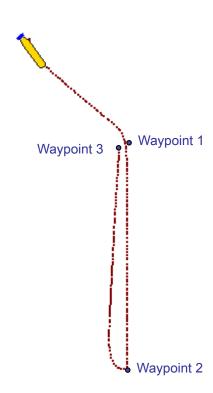
turn_range: Distance from a point, within which arrival is declared.

heading_avg = atan2(s, c) (180/ π)

$$s = \sum_{k=0}^{n-1} \sin(h_k \pi / 180)$$

$$c = \sum_{k=0}^{n-1} \cos(h_k \pi / 180)$$

turn radius: $r = v/((u/180)\pi)$







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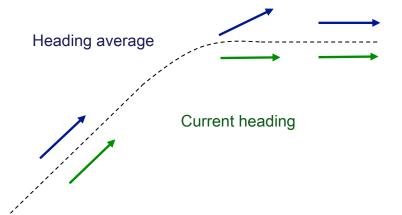
turn_range: Distance from a point, within which arrival is declared.

heading_avg = atan2(s, c) (180/ π)

$$s = \sum_{k=0}^{n-1} \sin(h_k \pi / 180)$$

$$c = \sum_{k=0}^{n-1} \cos(h_k \pi / 180)$$

turn radius: $r = v/((u/180)\pi)$







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Purpose: To limit the rate of vehicle turn to protect a towed sensor.

Parameters:

memory_time: A set of points in the X-Y plane, comprising a convex polygon.

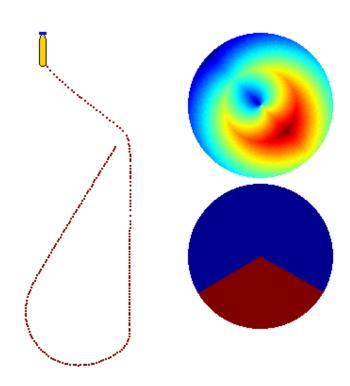
turn_range: Distance from a point, within which arrival is declared.

heading_avg = atan2(s, c) (180/ π)

$$s = \sum_{k=0}^{n-1} \sin(h_k \pi / 180)$$

$$c = \sum_{k=0}^{n-1} \cos(h_k \pi / 180)$$

turn radius: $r = v/((u/180)\pi)$







The Collision Avoidance Behavior

Purpose: Avoid collision with a vehicle with given position and trajectory.

Parameters:

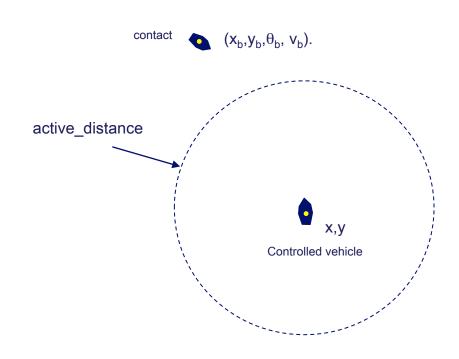
contact: Contact ID of the vehicle to be avoided.

active_distance: Distance to the contact beyond which the behavior has no influence on autonomy

all_clear_distance: CPA Distance to the contact below which penalty begins

collision distance: CPA Distance to the contact at which penalty is maximize (treated as a collision).

Utility based on CPA (closest point of approach)



Distance given by Pythagorean theorem in known trajectories. Express distance a function of angle, speed and time-on-leg.

$$\begin{split} \gamma^{2}(\theta, v, t) &= k_{2}t^{2} + k_{1}t + k_{0} \\ k_{2} &= \cos(\theta)v^{2} - 2\cos(\theta)v\cos(\theta_{b})v_{b} + \cos^{2}(\theta_{b})v_{b}^{2} + \\ &\sin^{2}(\theta)v_{2} - 2\sin(\theta)v\sin(\theta_{b})v_{b} + \sin^{2}(\theta_{b})v_{b}^{2} \\ k_{1} &= 2\cos(\theta)vy - 2\cos(\theta)vy_{b} - 2y\cos(\theta_{b})v_{b} + 2\cos(\theta_{b})v_{b}y_{b} + \\ &2\sin(\theta)vx - 2\sin(\theta)vx_{b} - \sin(\theta_{b})xv_{b} + 2\sin(\theta_{b})v_{b}x_{b} \\ k_{0} &= y_{2} - 2yy_{b} + y_{b}^{2} - 2xx_{b} + x_{b}^{2} \end{split}$$

First derivative w.r.t. time has only one root:

$$\gamma^{2}(\theta, v, t)' = 2k_{2}t + k_{1}$$
$$t = -\frac{k_{1}}{2k_{2}}$$





The Collision Avoidance Behavior

Purpose: Avoid collision with a vehicle with given position and trajectory.

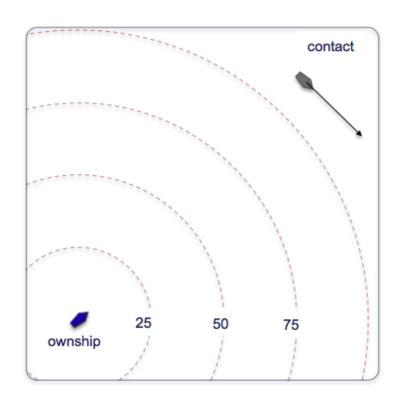
Parameters:

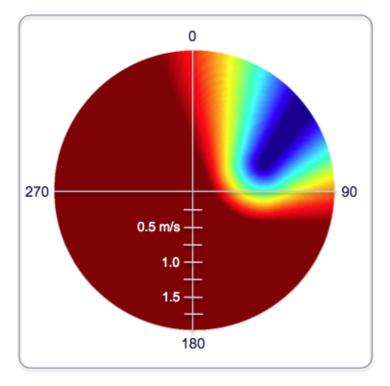
contact: Contact ID of the vehicle to be avoided.

active_distance: Distance to the contact beyond which the behavior has no influence on autonomy

all_clear_distance: CPA Distance to the contact below which penalty begins

collision distance: CPA Distance to the contact at which penalty is maximize (treated as a collision).









The Collision Avoidance Behavior

Purpose: Avoid collision with a vehicle with given position and trajectory.

Parameters:

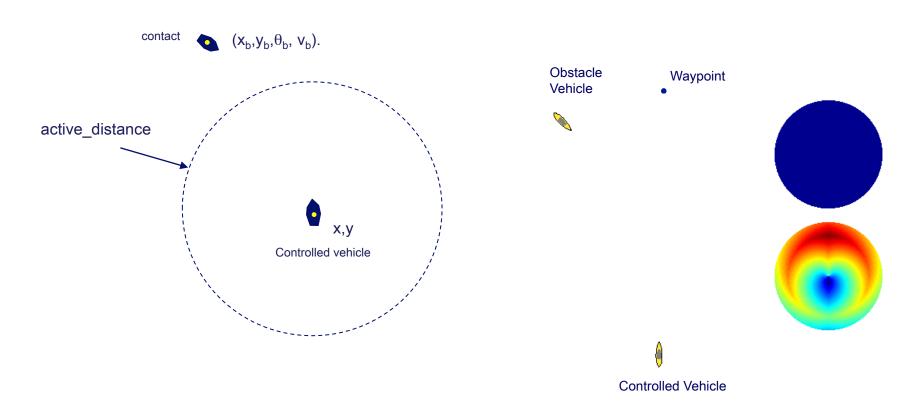
contact: Contact ID of the vehicle to be avoided.

active_distance: Distance to the contact beyond which the behavior has no influence on autonomy

all_clear_distance: CPA Distance to the contact below which penalty begins

collision distance: CPA Distance to the contact at which penalty is maximize (treated as a collision).

Utility based on CPA (closest point of approach)





Outline



- Trends in autonomous marine vehicles
- The Payload Autonomy Paradigm and the MOOS-IvP project
- Multi-Objective Optimization with Interval Programming
- ☐ The IvP Helm
- MOOS-IvP 4.2 and Plans for Future Development



MOOS-IvP 4.2



Changes to the Helm

- Preparation for multiple IvP Functions per behavior.
- Preparations for IvP Function re-use.
- Journaling of IVPHELM_STATUS messages (to reduce log file output)

New MOOS Apps

- uSimBeaconRange
- uSimActiveSonar
- uSimCurrent

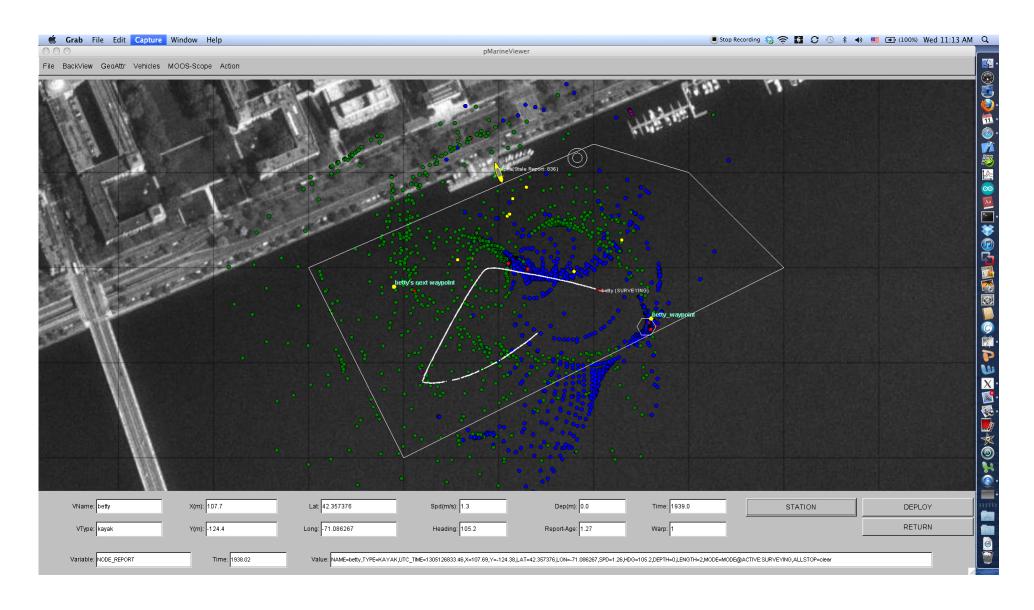
Improvements to Existing MOOS Apps

- pMarineViewer
- uSimMarine
- alogview
- uFunctionVis



MOOS-IvP 4.2





MONTEREY CALIFORNIA AUGUST 2006

Notable "First's":

- First usage of the payload autonomy paradigm
- First significant usage of MOOS-IvP on a UUV
- First deployment of a UUV with a vector sensor array

