

The Redstone Project

Satellite Network Control System Design

using MDP with Cooperative Game

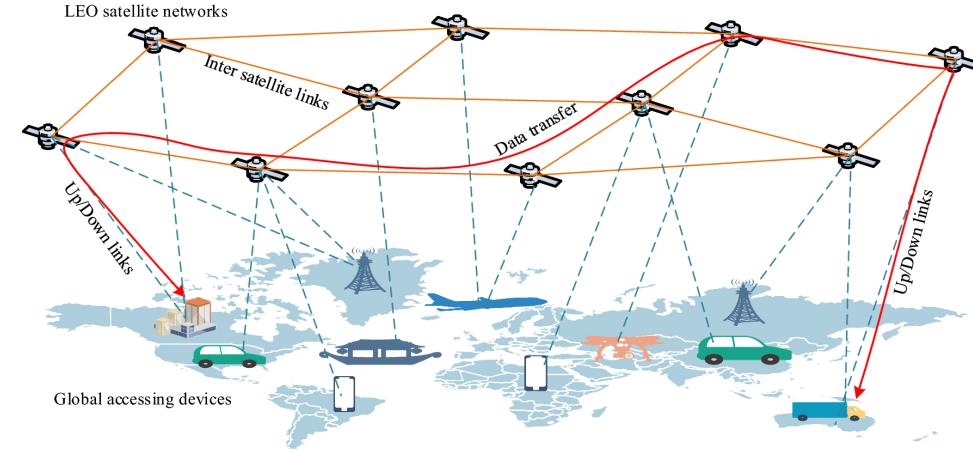


CENTER FOR
autonomy

Hongseok Kim
Oden Institute for Computational
Engineering and Sciences
08/03/2024

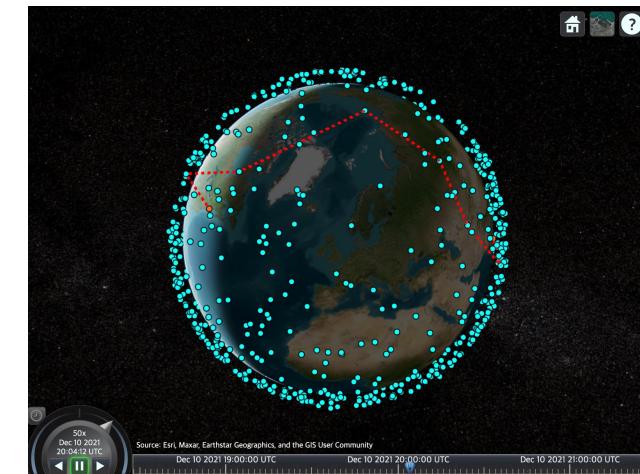
Problem Statement

LEO satellite networks



Global accessing devices

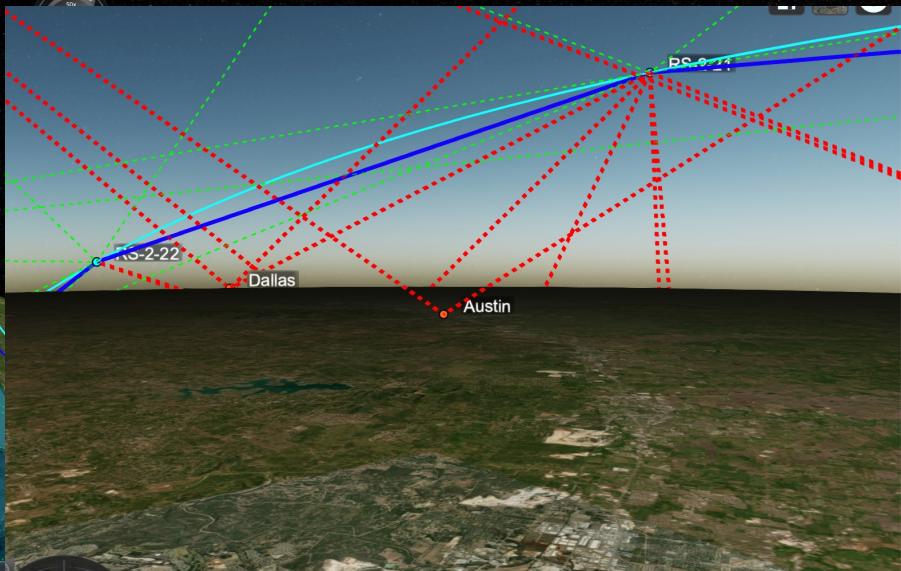
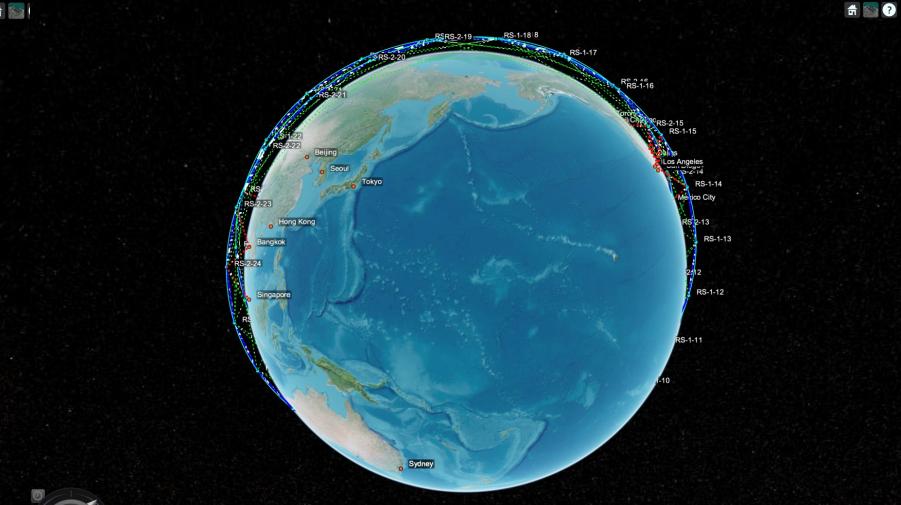
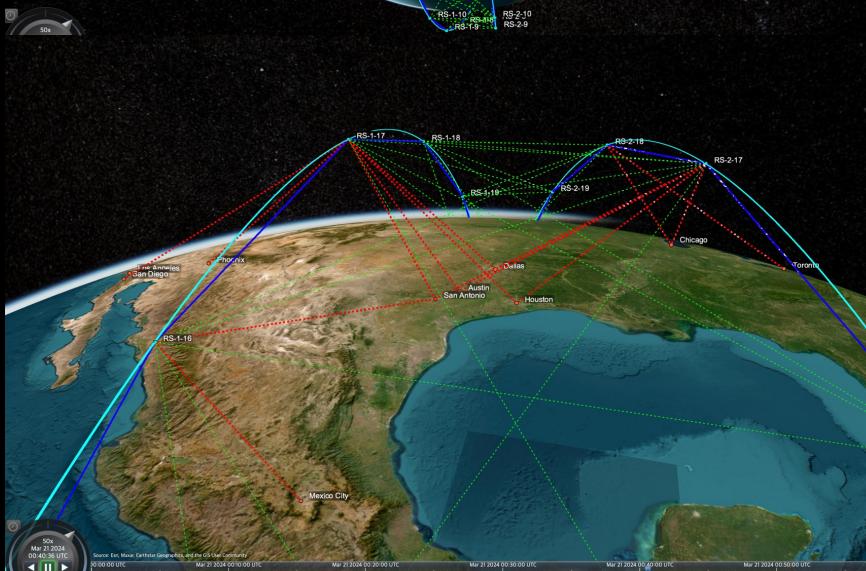
Zhao, N., Long, X. & Wang, J. A multi-constraint optimal routing algorithm in LEO satellite networks. *Wireless Netw* (2021).
<https://doi.org/10.1007/s11276-021-02674-3>



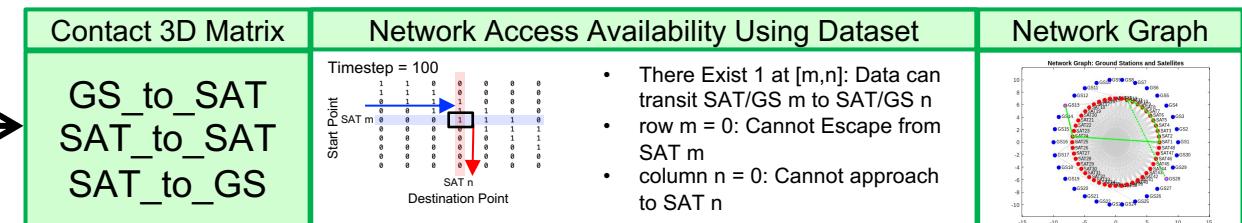
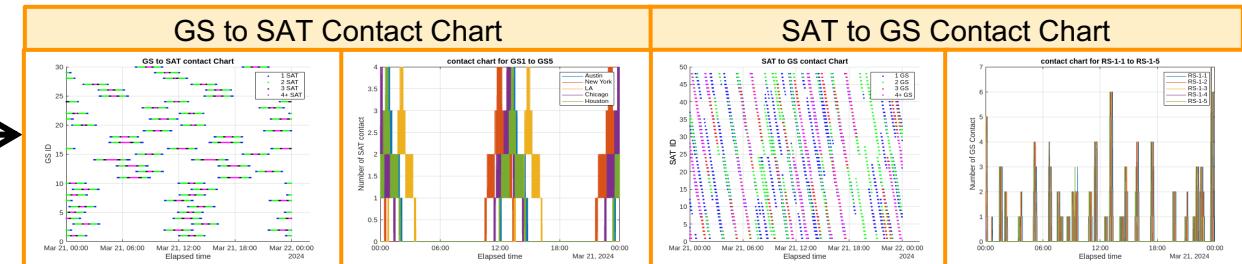
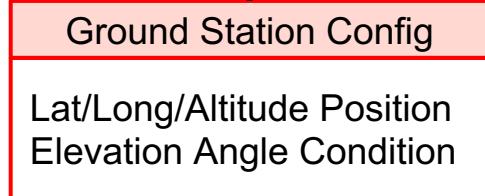
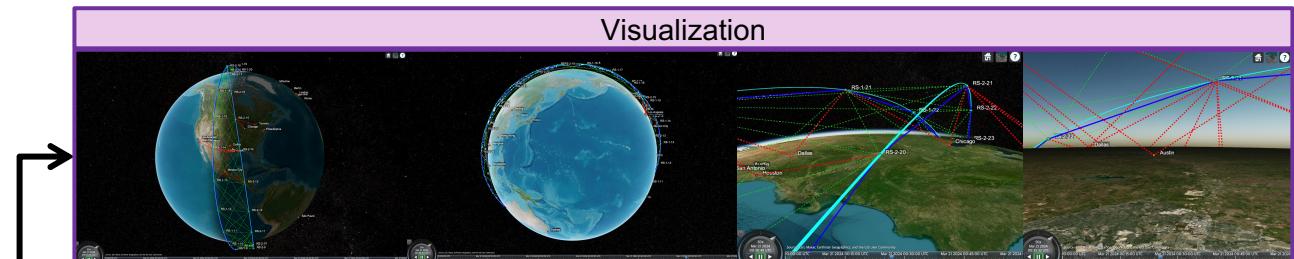
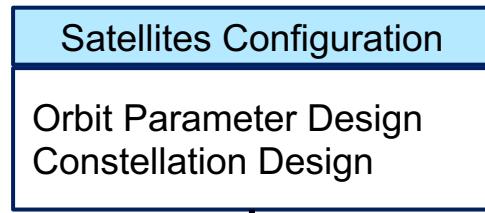
<https://www.mathworks.com/help/satcom/ug/multihop-path-select-through-sat-constellation.html>

I. Introduction

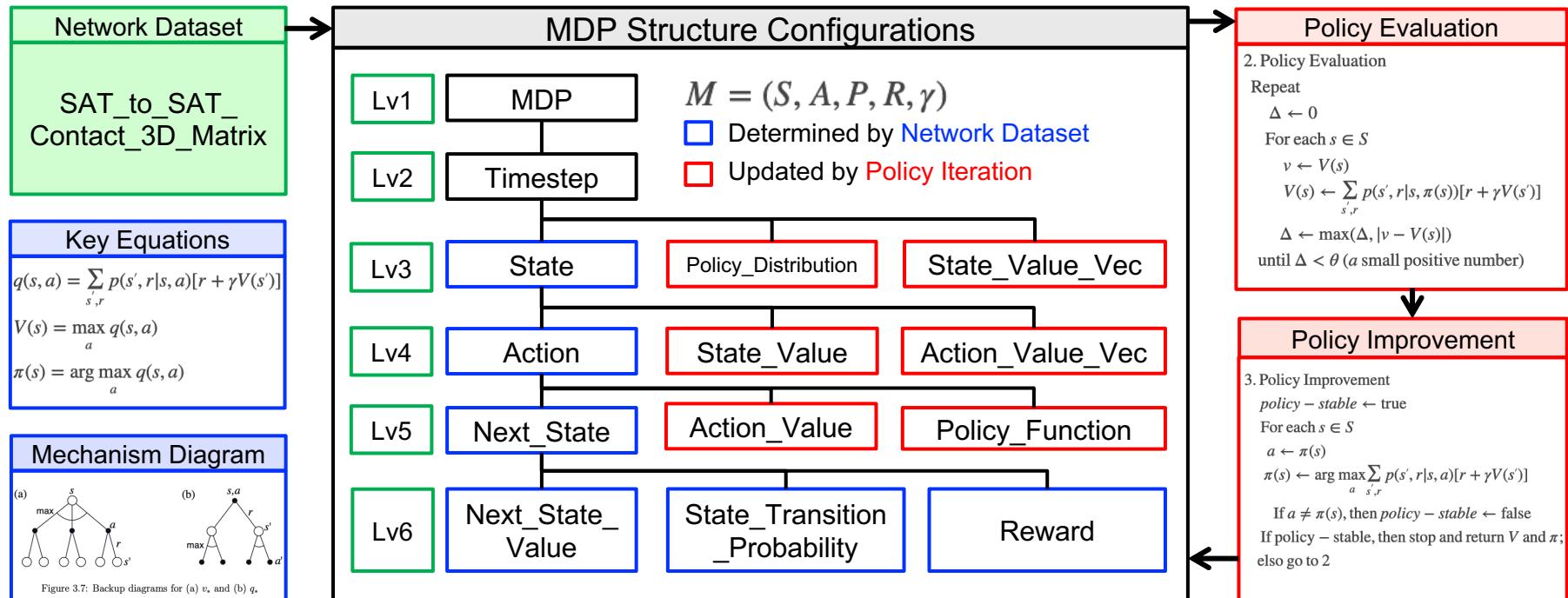
- Recently, ride-share launch vehicle service led lower cost of satellite operation, which made possible for design large-constellation for various missions.
- STARLINK currently operates 3000 satellites and OneWeb operates 700 satellite at LEO for low latency internet service globally.
- Planet Lab operates 300 satellites and South Korea ADD plans 100 satellites constellation at LEO for Earth Observation Mission, which minimizes re-visit time.
- Therefore, the demand for controlling constellation via autonomous network control sequence going to be higher, adapting optimized mission operation concept generated by machine learning solution.
- This project proposes the structure of satellite network control system and mission operation protocols in given operation requirements with simulation results.



II. Framework of the Project [1/3]



II. Framework of the Project [2/3]



II. Framework of the Project [3/3]

Time Variant , Multi Agent MDP Configuration

$$q(s, a, t) = \sum_{s', r} p(s', r|s, a) [r(s, a, s', t + 1) + \gamma V(s', t + 1)]$$

$$V(s, t) = \max_a q(s, a, t)$$

Indicates network change between t and t+1

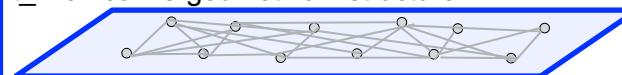
$$\pi(s, t) = \arg \max_a q(s, a, t)$$

Input info from $t : p(s', r|s, a) \leftarrow$ From Contact Matrix

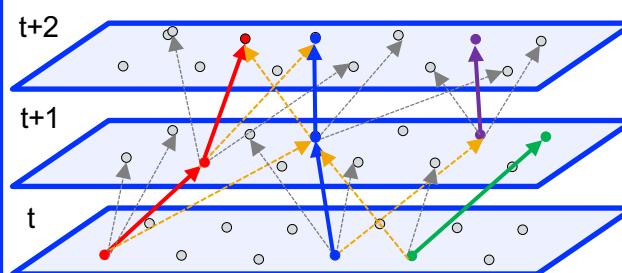
Input info from $t + 1 : V(s', t + 1), r(s, a, s', t + 1)$

Output info for $t : q(s, a, t), V(s, t), \pi(s, t)$

t_{final} : converged network structure



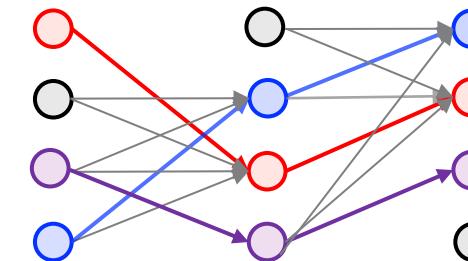
...



Data Packet Collision Avoidance Algorithm

Sequential Approach (Perfect Hierarchy)

Sequence:



: Actual Path by Agent
 : Deactivated after red and blue

`sat_to_sat(:,state_index,time_index-1) = 0;`

To avoid collision, after each agent's pathway propagation, deactivate every edge approaching to the node in which the agent stayed at given time

Parallel Approach (Cooperative Game)

Packet Collision Avoidance Algorithm – Cooperative Game

this example : active 4 states, each state have 4 available actions
 parse vector of active states $S_i = \{s_1, s_2, s_3, s_4\}$ at t_i

parse vector of available actions $a_i = \{a_{j1}, a_{j2}, a_{j3}, a_{j4}\}$

parse corresponding action values $q_i = \{q_{j1}, q_{j2}, q_{j3}, q_{j4}\}$

create set of action – action value vector from given state

\Rightarrow in state $s_j : \{(a_{j1}, q_{j1}), (a_{j2}, q_{j2}), (a_{j3}, q_{j3}), (a_{j4}, q_{j4})\}$

create matrix of states including these informations

$$\begin{bmatrix} s_1 & s_2 & s_3 & s_4 \\ (a_{11}, q_{11}) & (a_{21}, q_{21}) & (a_{31}, q_{31}) & (a_{41}, q_{41}) \\ (a_{12}, q_{12}) & (a_{22}, q_{22}) & (a_{32}, q_{32}) & (a_{42}, q_{42}) \\ (a_{13}, q_{13}) & (a_{23}, q_{23}) & (a_{33}, q_{33}) & (a_{43}, q_{43}) \\ (a_{14}, q_{14}) & (a_{24}, q_{24}) & (a_{34}, q_{34}) & (a_{44}, q_{44}) \end{bmatrix}$$

Caution : number of actions of each state are different

parse 1 action value from each active state

$$\begin{bmatrix} s_1 & s_2 & s_3 & s_4 & \Sigma q \\ \text{case 1} & q_{13} & q_{24} & q_{32} & q_{41} & Q_1 \\ \text{case 2} & q_{14} & q_{21} & q_{33} & q_{42} & Q_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \text{case 256} & q_{12} & q_{23} & q_{34} & q_{43} & Q_{256} \end{bmatrix} (Q_1 \geq Q_2 \geq \dots \geq Q_{256})$$

for case 1 : 256

if case k does not have collision

take case k : $\{(a_{13}, q_{13}), (a_{21}, q_{21}), (a_{34}, q_{34}), (a_{41}, q_{41})\}$

\Rightarrow collision resolved

III. Satellite Constellation Scenario Configuration [1/3]

3.1 Satellites Configuration

- Design Orbit: Kepler Orbit Elements

% Initial Orbit Settings

```
Altitude = 500; % (km)
Eccentricity = 1e-5;
Inclination = 97.4022; % (deg)
RAAN = 90; % (deg)
AOP = 0; % (deg)
% Number of satellites
number_of_SATs = 24;
```

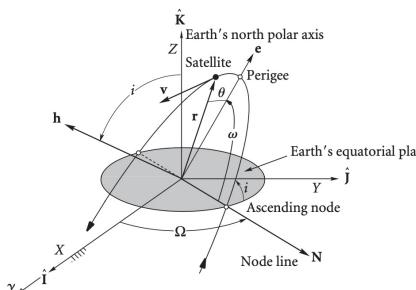
- Allocate Satellite in each orbit

```
% Satellite Characteristics - This will entail -1, -2, ... for
% constellation
satellite_name = "RS-1";

% Orbit Propagator
Orbit_Propagator = 'SGP4';

orbit_information.SMA = Altitude*1000 + 6378000;
orbit_information.ecc = Eccentricity;
orbit_information.inc = Inclination;
orbit_information.RAAN = RAAN;
orbit_information.AOP = AOP;

% Generate First Stream of Satellites
satellites_orbit_1 = single_orbit_constellation(scenario, ...
    orbit_information, ...
    number_of_SATs, ...
    Orbit_Propagator, ...
    satellite_name);
```



3.2 Ground Stations Configuration

- 3D Geographical Locations:
Latitude/Longitude/Altitude

```
gs_info = [
    "Austin", 30.2672, -97.7431, 15;
    "New York", 40.7128, -74.0068, 10;
    "Los Angeles", 34.0522, -118.4837, 71;
    "Chicago", 41.8781, -87.6298, 181;
    "Houston", 29.7604, -95.3698, 30;
    "Phoenix", 33.4484, -112.0740, 331;
    "Philadelphia", 39.9526, -75.1652, 12;
    "San Antonio", 29.4481, -98.4936, 198;
    "San Diego", 32.7157, -117.1611, 22;
    "Dallas", 32.7767, -96.7970, 137;
    "London", 51.5074, -0.1278, 35;
    "Tokyo", 35.6895, 139.6917, 6;
    "Paris", 48.8566, 2.3522, 35;
    "Moscow", 55.7558, 37.6176, 155;
    "Sydney", -33.8688, 151.2093, 19;
    "Toronto", 43.6511, -79.3830, 76;
    "Berlin", 52.5200, 13.4050, 34;
    "Rome", 41.9028, 12.4964, 17;
    "Seoul", 37.5665, 126.9780, 23;
    "Mumbai", 19.0760, 72.8777, 14;
    "Beijing", 39.9042, 116.4074, 43;
    "Mexico City", 19.4326, -99.1332, 2240];
```



- Elevation Angle Condition
(GS – SAT connectivity condition)

```
gs = groundStation(scenario, "Name", gs_info(:,1), ...
    "Latitude", str2double(gs_info(:,2)), ...
    "Longitude", str2double(gs_info(:,3)), ...
    "Altitude", str2double(gs_info(:,4)), ...
    "MinElevationAngle", 15);
```

III. Satellite Constellation Scenario Configuration [2/3]

3.3 Orbit Simulation Configuration

- Epoch / Timestep / Duration Initialization

```
% Initial Epoch % Duration and Steptime
Year = 2024; duration = hours(24);
Month = 3; steptime = 15; % seconds
Day = 21; startTime = datetime(Year,Month,Day,Hour,Minute,Seconds);
Hour = 0;
Minute = 0; stopTime = startTime + duration;
Seconds = 0; scenario = satelliteScenario(startTime,stopTime,steptime);
```

- SAT-to-SAT Access Matrix Generation

III. Create 3D inter-satellite contact (Diff orbit) matrix for each time step

```
sat1_to_sat2_contact_3d_matrix = zeros(24,24,length(time_vector));
sat2_to_sat1_contact_3d_matrix = zeros(24,24,length(time_vector));

for i = 1:24
    sat1_to_sat2_access_status = accessStatus(access(satellites_orbit_1(i),satellites_orbit_2));
    sat1_to_sat2_contact_3d_matrix(i,:,:) = sat1_to_sat2_access_status;
    sat2_to_sat1_contact_3d_matrix(:,:,:) = sat1_to_sat2_access_status;
end
```

IV. Create 3D inter-satellite contact (Same orbit) matrix for each time step

```
sat1_to_sat1_contact_3d_matrix = zeros(24,24,length(time_vector));

sat1_to_sat1_contact_3d_matrix(1,1,:) = 1;
sat1_to_sat1_contact_3d_matrix(2,1,:) = 1;
sat1_to_sat1_contact_3d_matrix(1,2,:) = 1;
sat1_to_sat1_contact_3d_matrix(24,1,:) = 1;
sat1_to_sat1_contact_3d_matrix(1,24,:) = 1;
sat1_to_sat1_contact_3d_matrix(24,23,:) = 1;
sat1_to_sat1_contact_3d_matrix(23,24,:) = 1;
sat1_to_sat1_contact_3d_matrix(24,24,:) = 1;

for i = 2:23
    sat1_to_sat1_contact_3d_matrix(i,i,:) = 1;
    sat1_to_sat1_contact_3d_matrix(i+1,i,:) = 1;
    sat1_to_sat1_contact_3d_matrix(i,i+1,:) = 1;
    sat1_to_sat1_contact_3d_matrix(i,-1,:) = 1;
    sat1_to_sat1_contact_3d_matrix(-1,i,:) = 1;
end

sat2_to_sat2_contact_3d_matrix = sat1_to_sat1_contact_3d_matrix;
sat_to_sat_contact_3d_matrix = [sat1_to_sat1_contact_3d_matrix, sat2_to_sat1_contact_3d_matrix;
    sat1_to_sat2_contact_3d_matrix, sat2_to_sat2_contact_3d_matrix];
```

- GS – to - SAT Access Matrix Generation

II. Create 3D ground-to-satellite contact matrix for each time step

```
number_of_gs =length(gs);
number_of_SATs = length(satellites);
time_vector = scenario.StartTime:seconds(scenario.SampleTime):scenario.StopTime;

gs_to_sat_contact_3d_matrix = zeros(number_of_gs, number_of_SATs, length(time_vector));

for i = 1:number_of_gs
    gs_to_sat_access_Status = accessStatus(access(gs(i), satellites));
    gs_to_sat_contact_3d_matrix(i,:,:,:) = gs_to_sat_access_Status;
end
```

- Resultant Contact Table

gs_to_sat_access_table = 273x8 table

	Source	Target	IntervalNumber	StartTime	EndTime	Duration
1	"Austin"	"RS-1-1"	1	2024-03-21 00:05:15	2024-03-21 00:11:00	345
2	"Austin"	"RS-1-2"	1	2024-03-21 00:01:30	2024-03-21 00:07:00	330
3	"Austin"	"RS-1-3"	1	2024-03-21 00:00:00	2024-03-21 00:03:00	180
4	"Austin"	"RS-1-12"	1	2024-03-21 00:57:00	2024-03-21 01:00:00	180
5	"Austin"	"RS-1-13"	1	2024-03-21 00:53:00	2024-03-21 00:57:15	255
6	"Austin"	"RS-1-14"	1	2024-03-21 00:48:45	2024-03-21 00:53:30	285
7	"Austin"	"RS-1-15"	1	2024-03-21 00:44:45	2024-03-21 00:49:45	300
8	"Austin"	"RS-1-16"	1	2024-03-21 00:40:45	2024-03-21 00:45:45	300
9	"Austin"	"RS-1-17"	1	2024-03-21 00:36:45	2024-03-21 00:42:00	315

sat_to_gs_access_table = 273x8 table

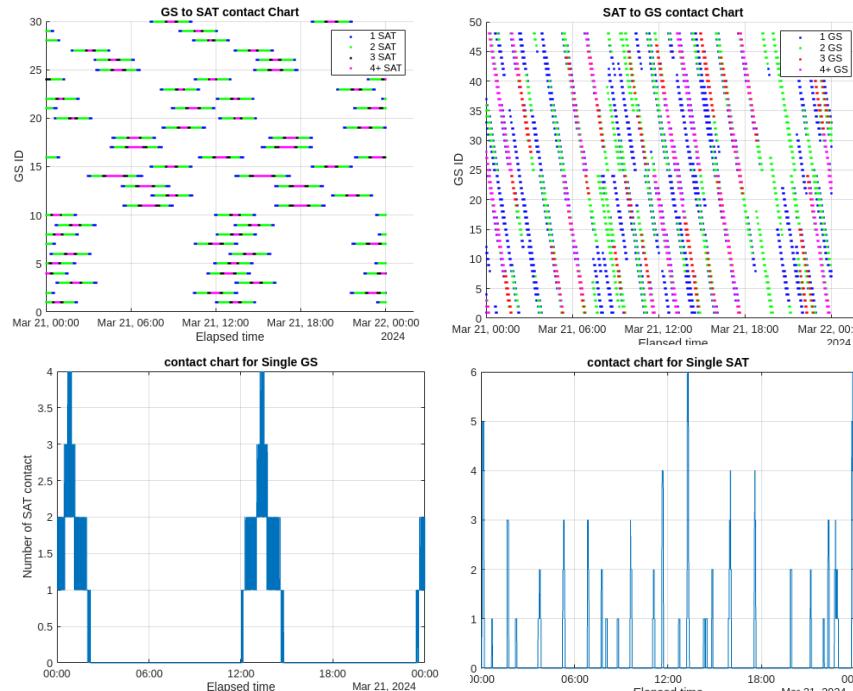
	Source	Target	IntervalNumber	StartTime	EndTime	Duration
1	"RS-1-1"	"Austin"	1	2024-03-21 00:05:15	2024-03-21 00:11:00	345
2	"RS-1-1"	"Chicago"	1	2024-03-21 00:08:45	2024-03-21 00:13:00	255
3	"RS-1-1"	"Houston"	1	2024-03-21 00:05:00	2024-03-21 00:10:45	345
4	"RS-1-1"	"San Antonio"	1	2024-03-21 00:05:15	2024-03-21 00:10:45	330
5	"RS-1-1"	"Dallas"	1	2024-03-21 00:06:00	2024-03-21 00:11:30	330
6	"RS-1-1"	"Mumbai"	1	2024-03-21 00:41:15	2024-03-21 00:44:15	180
7	"RS-1-1"	"Mexico City"	1	2024-03-21 00:03:00	2024-03-21 00:08:00	300
8	"RS-1-2"	"Austin"	1	2024-03-21 00:01:30	2024-03-21 00:07:00	330
9	"RS-1-2"	"Chicago"	1	2024-03-21 00:04:30	2024-03-21 00:09:15	285

III. Satellite Constellation Scenario Configuration [3/3]

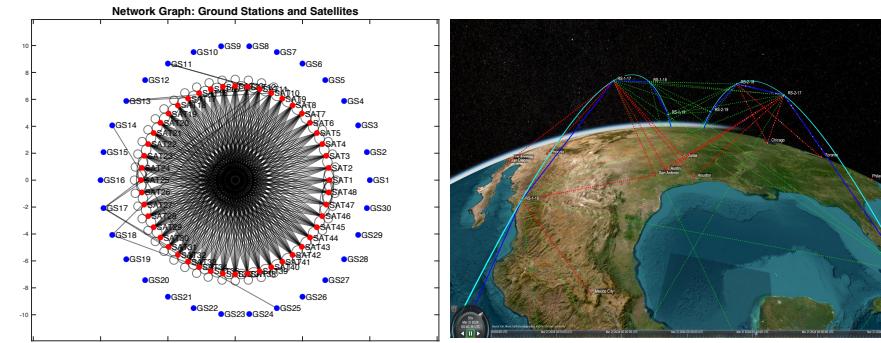
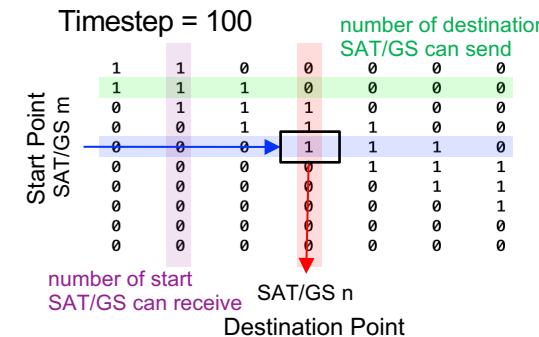
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3.4 Orbit Simulation result

- GS → SAT / SAT → GS contact chart generation



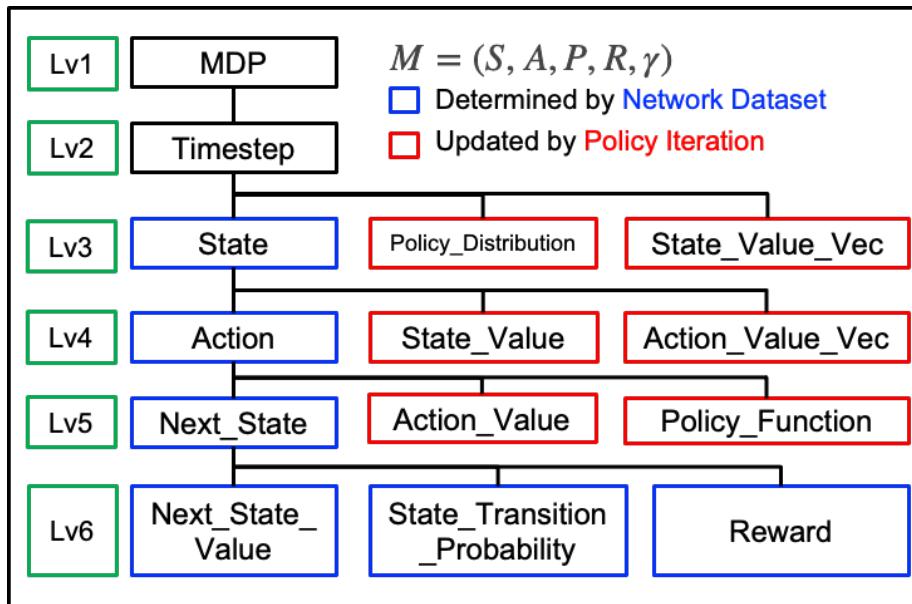
- Contact 3D Matrix -> network graph generation



IV. MDP-DP structure Configuration [1/3]

4.1 Network Dataset to MDP Structure Definition

- In creating MDP framework environment, we need 5 necessary variable space $M = (S, A, P, R, \gamma)$ and policy $\pi(a | s)$
- Since we are simulating the network structure, we define action as transiting one data packet from state to state
- Caution: The satellites and ground stations themselves are not the states. The phenomenon that each active data packet is located at each agent are states.**



S : Finite Set of States A : Finite Set of Actions γ : Discount Factor
 P : State Transition Probability Matrix R : Reward Function
 Action Value : $q(s, a) = \sum_{s', r} p(s', r | s, a)[r(s, a, s') + \gamma V(s')]$
 State Value : $V(s) = \max q(s, a)$
 Policy Function : $\pi(s|a) = \arg \max_a q(s, a, t)$

(a)

Diagram illustrating the calculation of $q(s, a)$ for a state s . The root node is s , which has three children. The top child is labeled $V(s)$ and $\pi(a | s)$. The middle child is labeled a and $q(s, a)$. The bottom child is labeled $r(s, a, s')$ and $s' V(s')$. A red bracket on the right indicates the formula $p(s', r | s, a)$.

Question:
 How to get action value from state-action space by given policy?

IV. MDP-DP structure Configuration [2/3]

4.2 Using Dynamic Programming (Policy Iteration) for solving MDP

Policy Iteration (Using iterative policy evaluation)

1. Initialization

$V(s) \in R$ and $\pi(s) \in A(s)$ arbitrarily for all $s \in S$

2. Policy Evaluation

Repeat

$$\Delta \leftarrow 0$$

For each $s \in S$

$$v \leftarrow V(s)$$

$$V(s) \leftarrow \sum_{s',r} p(s',r|s,\pi(s))[r + \gamma V(s')]$$

$$\Delta \leftarrow \max(\Delta, |v - V(s)|)$$

until $\Delta < \theta$ (θ a small positive number)

3. Policy Improvement

$policy-stable \leftarrow true$

For each $s \in S$

$updated \leftarrow \pi(s)$

$$\pi(s) \leftarrow \arg \max_{a,s',r} \sum_{s',r} p(s',r|s,a)[r + \gamma V(s')]$$

If $a \neq \pi(s)$, then $policy-stable \leftarrow false$

If $policy-stable$, then stop and return V and π ;
 else go to 2

Reward System by next state by different cases

Case 1: Regular transition

- Currently, we have two sub-orbits of the satellite system: Orbit 1 and Orbit 2.
- Transiting between different orbit may take more cost, which means less reward.

1-1: From orbit 1 to orbit 1: -1

1-2: From orbit 1 to orbit 2: -15

1-3: From orbit 2 to orbit 1: -1

1-4: From orbit 2 to orbit 2: -15

1-5: Returning to starting state: -15

Case 2: Next State = end state

2-1: From orbit 1 to orbit 1: 100

2-2: From orbit 1 to orbit 2: 50

2-3: From orbit 2 to orbit 1: 100

2-4: From orbit 2 to orbit 2: 50

2-5: Returning to starting state: -15

Case 3: Starting State = end state

3-1: Returing to starting state: 0

simulation set up complete!

Policy: 1 \rightarrow Value Iteration: 466

Policy: 2 \rightarrow Value Iteration: 14

Policy: 3 \rightarrow Value Iteration: 5

Policy: 4 \rightarrow Value Iteration: 5

Policy: 5 \rightarrow Value Iteration: 8

Policy: 6 \rightarrow Value Iteration: 10

Policy: 7 \rightarrow Value Iteration: 9

Policy: 8 \rightarrow Value Iteration: 1

Policy: 9 \rightarrow Value Iteration: 1

Policy: 10 \rightarrow Value Iteration: 1

Policy: 11 \rightarrow Value Iteration: 1

v_k for the Random Policy

0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0

Greedy Policy w.r.t. v_k

■	■	■	■
■	■	■	■
■	■	■	■
■	■	■	■
■	■	■	■

random policy

$k=0$

0.0	-1.0	-1.0	-1.0
-1.0	-1.0	-1.0	-1.0
-1.0	-1.0	-1.0	-1.0
-1.0	-1.0	-1.0	-1.0
-1.0	-1.0	-1.0	-1.0

$k=1$

0.0	-1.7	-2.0	-2.0
-1.7	-2.0	-2.0	-2.0
-2.0	-2.0	-2.0	-1.7
-2.0	-2.0	-1.7	0.0
-2.0	-2.0	-1.7	0.0

$k=2$

0.0	-2.4	-2.9	-3.0
-2.4	-2.9	-3.0	-2.0
-2.9	-3.0	-2.9	-2.4
-3.0	-2.9	-2.4	0.0
-3.0	-2.9	-2.4	0.0

$k=3$

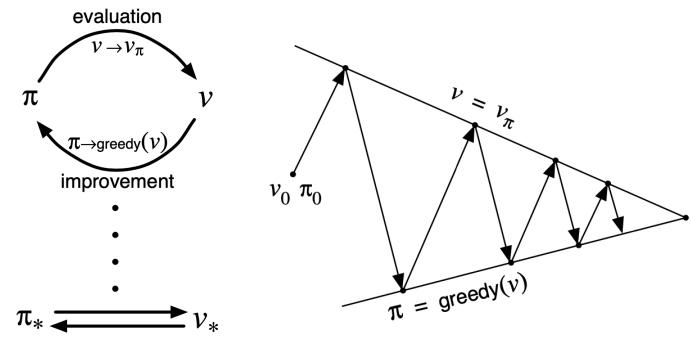
0.0	-6.1	8.4	-9.0
-6.1	-7.7	8.4	-8.4
-8.4	8.4	7.7	-6.1
-9.0	8.4	6.1	0.0
-9.0	8.4	6.1	0.0

$k=10$

0.0	-14	-20	-22
-14	-18	-20	-20
-20	-20	-18	-14
-22	-20	-14	0.0
-22	-20	-14	0.0

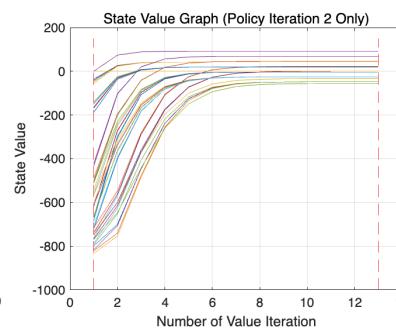
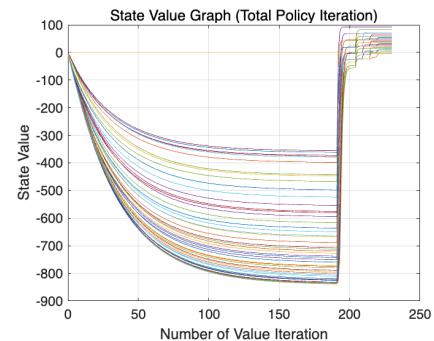
$k=\infty$

0.0	-14	-20	-22
-14	-18	-20	-20
-20	-20	-18	-14
-22	-20	-14	0.0
-22	-20	-14	0.0

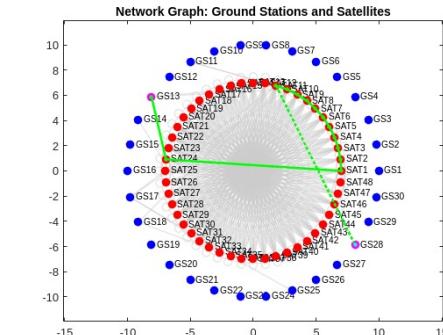
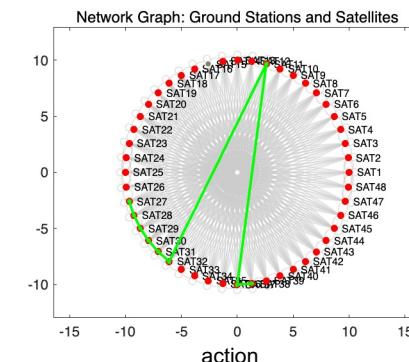


IV. MDP-DP structure Configuration [3/3]

4.3 State Value Change by Value/Policy Iteration



4.4 Generated Network Path (Time Invariant)



Simulation Result

state_list	cumulative_reward_list	state_value_list
3	0	2
4	-1	10.5
5	-2	19
31	-17	41.5
32	-18	50
33	-19	58.5
34	-20	67
35	-21	75.5
36	-22	84
37	-23	92.5
38	77	0

← Policy Distribution

V. Time Variant – Multi Agent MDP Configuration [1/2]

5.1 Basic Mechanism – time propagation

$$q(s, a, t) = \sum_{s', r} p(s', r|s, a)[r(s, a, s', t+1) + \gamma V(s', t+1)]$$

$$V(s, t) = \max_a q(s, a, t)$$

Indicates network change between t and t+1

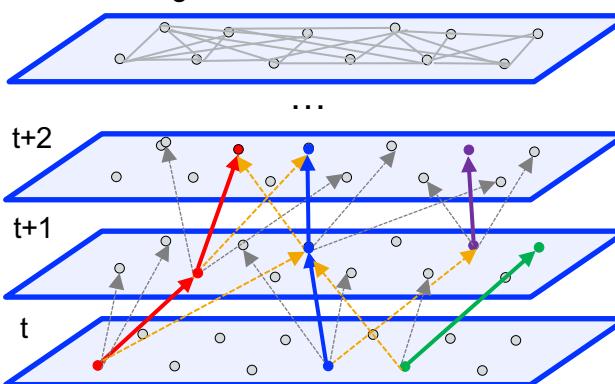
$$\pi(s, t) = \arg \max_a q(s, a, t)$$

Input info from t : $p(s', r|s, a) \leftarrow$ From Contact Matrix

Input info from $t+1$: $V(s', t+1), r(s, a, s', t+1)$

Output info for t : $q(s, a, t), V(s, t), \pi(s, t)$

t final: converged network structure



Steps for Time Variant MDP – DP

Step 1 : Policy Iteration for $T = T_0 + \Delta T$

Output info for the input : $V(s, T_0 + \Delta T)$

Step 2 : State Value Propagation from $T = T_0 + \Delta T$ to $T = T_0$

$$q(s, a, t) = \sum_{s', r} p(s', r|s, a)[r(s') + \gamma V(s', t+1)] \rightarrow \text{Level 4}$$

$$V(s, t) = \max_a \sum_{s', r} p(s', r|s, a)[r(s') + \gamma V(s', t+1)] \rightarrow \text{Level 3}$$

Input info from $T = t$: $p(s', r|s, a) \Rightarrow$ From Satellite to Satellite Contact Matrix at $T = t$

– We have to define action from $s(t)$ to $s'(t+1)$ and corresponding rewards

Input info from $T = t+1$: $V(s', t+1) \Rightarrow$ From State value (level 3) at $T = t+1$

Output info for $T = t$: $V(s, t), q(s, a, t) \Rightarrow$ Reuse this information to calculate $V(s, t-1), q(s, a, t-1)$

Step 3 : Policy Determination Based of State Value

For each $T = t$,

$$\pi(s, t) = \arg \max_a \sum_{s', r} p(s', r|s, a)[r(s') + \gamma V(s', t+1)]$$

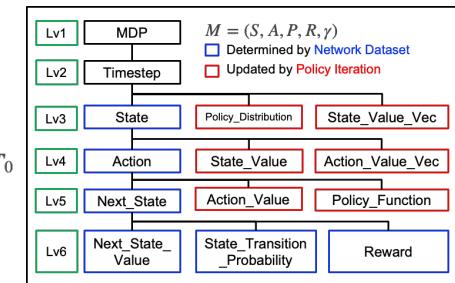
If there is n multiple actions which have same action value, $\pi = \frac{1}{n}$ for each action

Step 4 : Test the simulation

Input : S_n at $T = T_0$

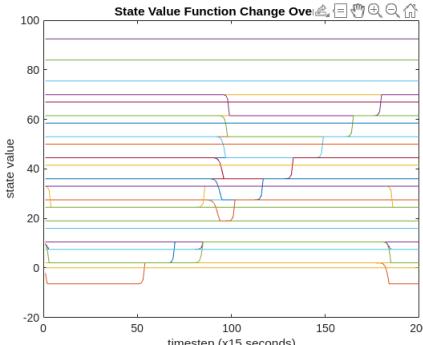
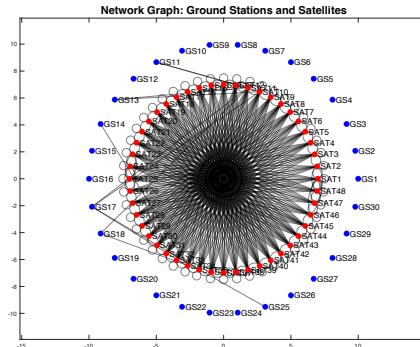
Output : S_m at $T = T_0 + \Delta T$

Or any time, any input can be possible, check whether the output is intended destination

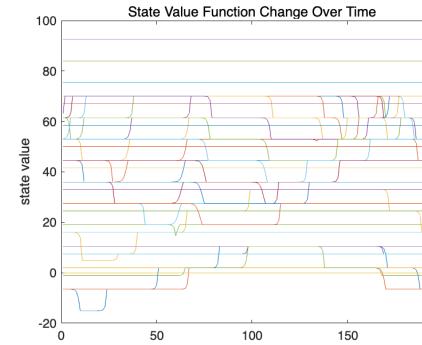
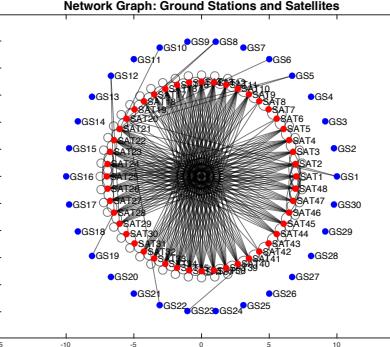


V. Time Variant – Multi Agent MDP Configuration [2/2]

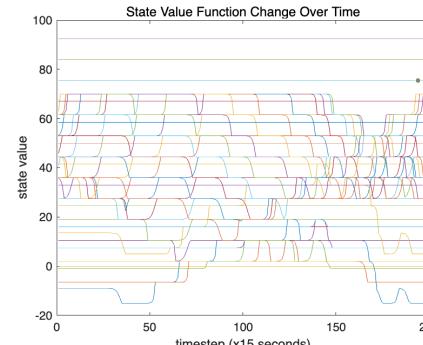
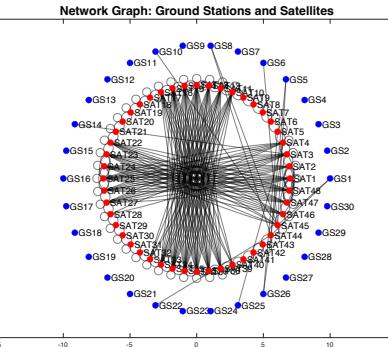
5.2 State Value Distribution Change over time: reflects network change



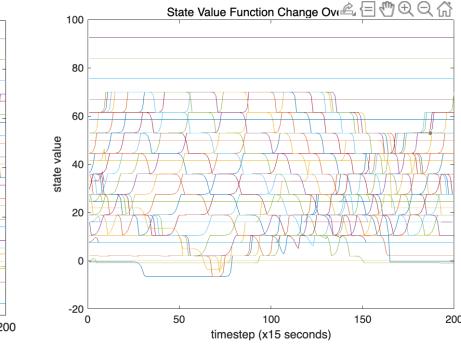
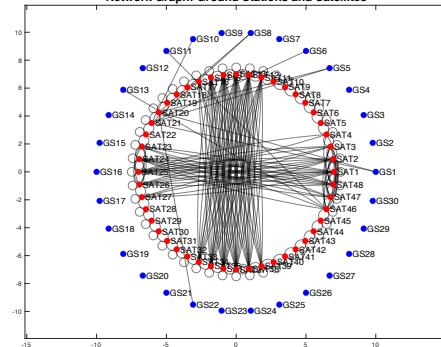
RAAN difference = 15 deg



RAAN difference = 45 deg



RAAN difference = 60 deg

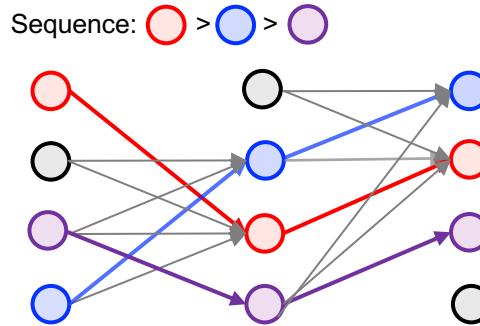


RAAN difference = 90 deg

VI. Collision Avoidance Algorithm [1/4]

6.1 Sequential Approach (Perfect Hierarchy)

- Algorithm Explanation, Code Configuration

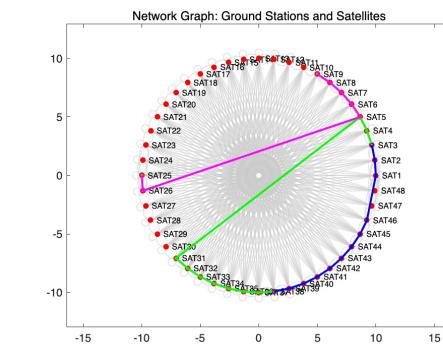
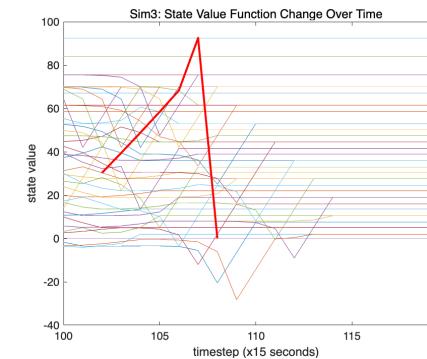
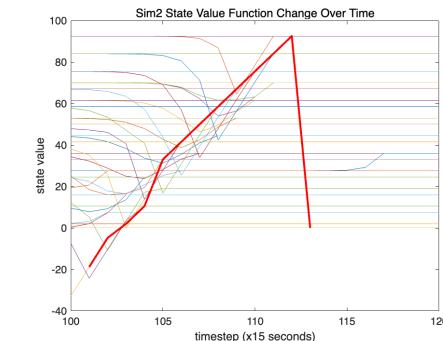
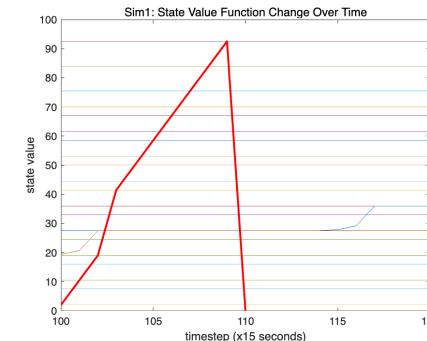


: Actual Path by Agent
 : Deactivated after red and blue

`sat_to_sat(:,state_index,time_index-1) = 0;`

To avoid collision, after each agent's pathway propagation, deactivate every edge approaching to the node in which the agent stayed at given time

- State Value change by every path



VI. Collision Avoidance Algorithm [2/4]

6.2 Parallel Approach (Cooperative Game)

Packet Collision Avoidance Algorithm – Cooperative Game

this example : active 4 states, each state have 4 available actions

parse vector of active states $S_t = \{s_1, s_2, s_3, s_4\}$ at t_i

parse vector of available actions $a_i \equiv \{a_{i1}, a_{i2}, a_{i3}, a_{i4}\}$

parse corresponding action values $a_i = \{a_{i1}, a_{i2}, a_{i3}, a_{i4}\}$

Create set of action – action value vector from given state

\Rightarrow in state $s_i : \{(q_{i1}, q_{i1}), (q_{i2}, q_{i2}), (q_{i3}, q_{i3}), (q_{i4}, q_{i4})\}$

$$\Rightarrow \begin{bmatrix} s_1 & s_2 & s_3 & s_4 \\ (a_{11}, q_{11}) & (a_{21}, q_{21}) & (a_{31}, q_{31}) & (a_{41}, q_{41}) \\ (a_{12}, q_{12}) & (a_{22}, q_{22}) & (a_{32}, q_{32}) & (a_{42}, q_{42}) \\ (a_{13}, q_{13}) & (a_{23}, q_{23}) & (a_{33}, q_{33}) & (a_{43}, q_{43}) \\ (a_{14}, q_{14}) & (a_{24}, q_{24}) & (a_{34}, q_{34}) & (a_{44}, q_{44}) \end{bmatrix}$$

Caution : number of actions of each state are different

parse 1 action value from each active state

$$\Rightarrow \begin{bmatrix} & s_1 & s_2 & s_3 & s_4 & \Sigma q \\ \text{case 1} & q_{13} & q_{24} & q_{32} & q_{41} & Q_1 \\ \text{case 2} & q_{14} & q_{21} & q_{33} & q_{42} & Q_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \text{case256} & q_{12} & q_{23} & q_{34} & q_{43} & Q_{256} \end{bmatrix} \quad (Q_1 \geq Q_2 \geq \dots \geq Q_{256})$$

for case 1 : 256

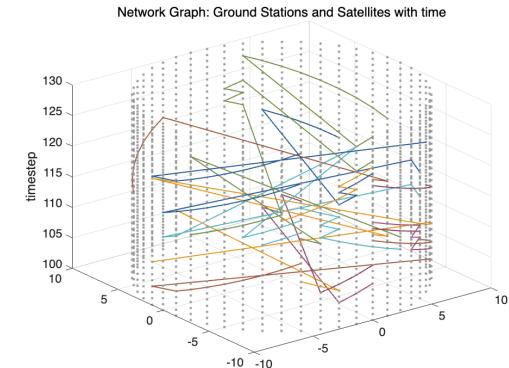
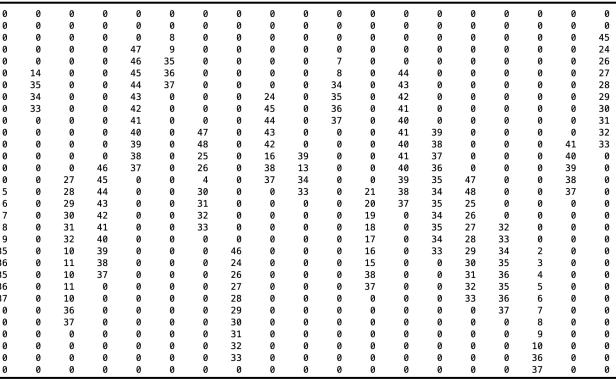
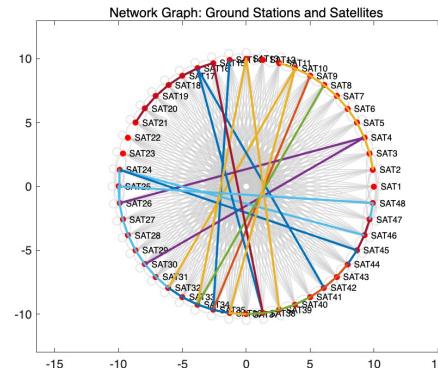
if case k does not have collision

take case k : $\{ (a_{13}, q_{13}), (a_{21}, q_{21}), (a_{34}, q_{34}), (a_{41}, q_{41}) \}$

⇒ collision resolved

```
simulation set up complete!
Policy: 1 -> Value Iteration: 41
Policy: 2 -> Value Iteration: 14
Policy: 3 -> Value Iteration: 5
Policy: 4 -> Value Iteration: 5
Policy: 5 -> Value Iteration: 5
Policy: 6 -> Value Iteration: 10
Policy: 7 -> Value Iteration: 9
Policy: 8 -> Value Iteration: 1
Policy: 9 -> Value Iteration: 1
```

```
collision occurred at time index 109  
collision resolved at time index 109 at case index  
collision occurred at time index 111  
collision resolved at time index 111 at case index  
collision occurred at time index 112  
collision resolved at time index 112 at case index  
collision occurred at time index 113  
collision resolved at time index 113 at case index
```



VI. Collision Avoidance Algorithm [3/4]

6.3 Comparison between two methodology

- Pros and Cons
- Quantified Performance Comparison
- state Value / Action Value distribution by each agent
- Total Reward change over time by each agent

```

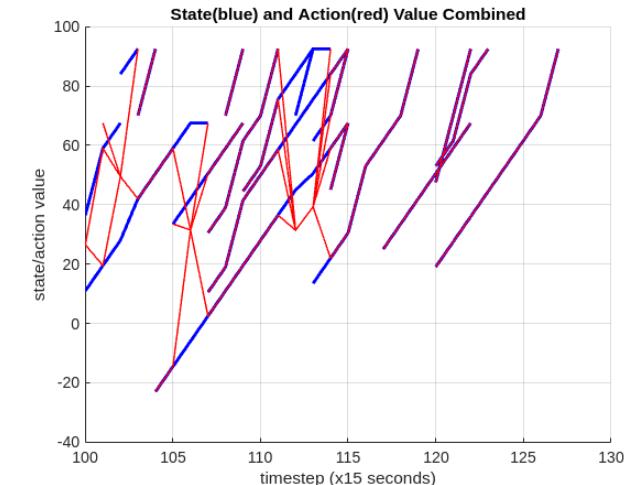
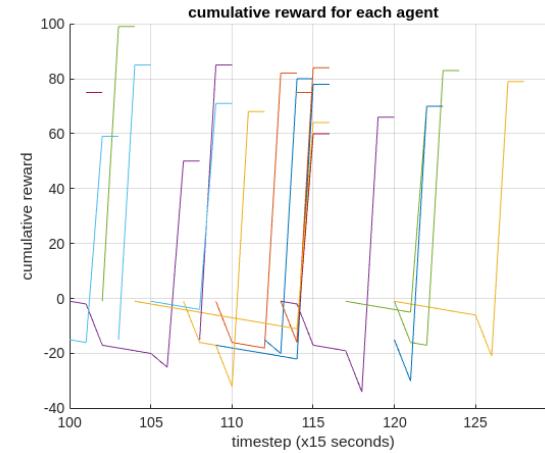
collision occurred at time index 102
No. of Active Agents: 4, No. of Agent-Action Cases: 64
Optimal Action Values:
84.00 0.00 28.00 67.50
Sum: 179.50
Updated Action Values at case index 2:
84.00 0.00 21.20 67.50
Sum: 172.70
Collision Resolved!

```

```

collision occurred at time index 106
No. of Active Agents: 3, No. of Agent-Action Cases: 64
Optimal Action Values:
42.00 -6.00 67.50
Sum: 103.50
Updated Action Values at case index 5:
42.00 -6.00 57.50
Sum: 93.50
Collision Resolved!

```



VI. Collision Avoidance Algorithm [4/4]

6.4 Using Congestion Game

Packet Collision Avoidance Algorithm – Congestion Game

this example : active 4 states, each state have 4 available actions

parse vector of active states $S_t = \{s_1, s_2, s_3, s_4\}$ at t_i

parse vector of available actions $a_i = \{a_{j1}, a_{j2}, a_{j3}, a_{j4}\}$

parse corresponding action values $q_i = \{q_{j1}, q_{j2}, q_{j3}, q_{j4}\}$

parse corresponding next state values $s'_i = \{s'_{j1}, s'_{j2}, s'_{j3}, s'_{j4}\}$

create set of action – action value – next state vector from given state

\Rightarrow in state $s_j : \{(a_{j1}, q_{j1}, s'_{j1}), (a_{j2}, q_{j2}, s'_{j2}), (a_{j3}, q_{j3}, s'_{j3}), (a_{j4}, q_{j4}, s'_{j4})\}$

create dataset of current states including these informations

$$\text{SAS}_t = \begin{bmatrix} s_1 - \pi_1 & s_2 - \pi_2 & s_3 - \pi_3 & s_4 - \pi_4 \\ (a_{11}, q_{11}, s'_{11}) & (a_{21}, q_{21}, s'_{21}) & (a_{31}, q_{31}, s'_{31}) & (a_{41}, q_{41}, s'_{41}) \\ (a_{12}, q_{12}, s'_{12}) & (a_{22}, q_{22}, s'_{22}) & (a_{32}, q_{32}, s'_{32}) & (a_{42}, q_{42}, s'_{42}) \\ (a_{13}, q_{13}, s'_{13}) & (a_{23}, q_{23}, s'_{23}) & (a_{33}, q_{33}, s'_{33}) & (a_{43}, q_{43}, s'_{43}) \\ (a_{14}, q_{14}, s'_{14}) & (a_{24}, q_{24}, s'_{24}) & (a_{34}, q_{34}, s'_{34}) & (a_{44}, q_{44}, s'_{44}) \end{bmatrix}$$

while true (infinite loop)

parse current policy distribution for each active state

policy = $[\pi_1(a_{11}|s_1) = 1 \ \ \pi_2(a_{21}|s_2) = 1 \ \ \pi_3(a_{31}|s_3) = 1 \ \ \pi_4(a_{41}|s_4) = 1]$

caution : number of actions of each state are different

create dataset of next state applying current policy π for each state

\rightarrow Newly Generated States may be included in this case

$$\text{SAS}_{t+1} = \begin{bmatrix} s_1 & s_2 & s_3 & s_4 & s_5 \\ s'_{11} & s'_{21} & s'_{31} & s'_{41} & s'_{51} \end{bmatrix} \quad (s_5 \text{ is newly generated packet at } t+1)$$

for active states $S_t = \{s_1, s_2, s_3, s_4\}$

find number of colliding states for all allocation for current state

$s'_{j1} \dots s'_{j4} \in \{s'_{21}, s'_{31}, s'_{41}, s'_{51}\}$

update action values for each action by applying penalty from collision

if s'_{j1} has k collisions $\Rightarrow q'_{j1} = q_{j1} - ka$ (a : Collision Factor)

update policy by finding maximum action values updated

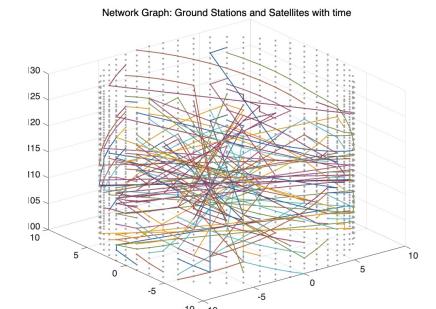
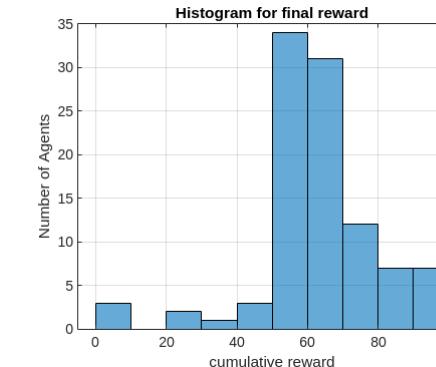
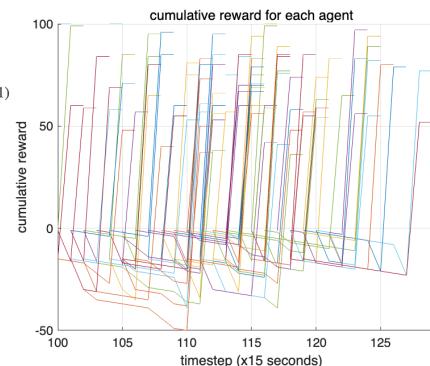
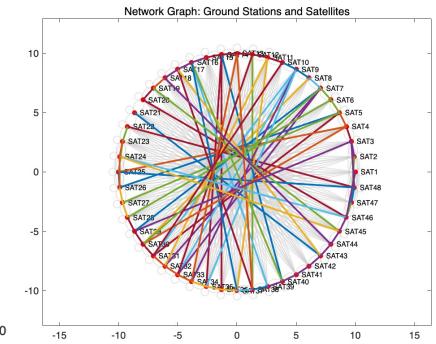
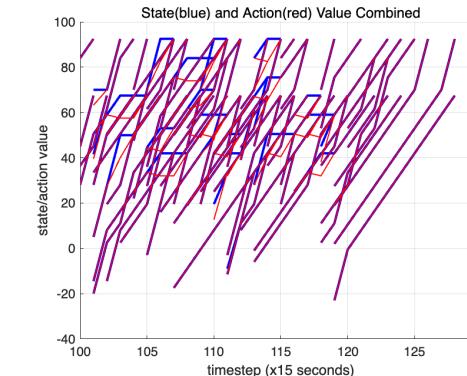
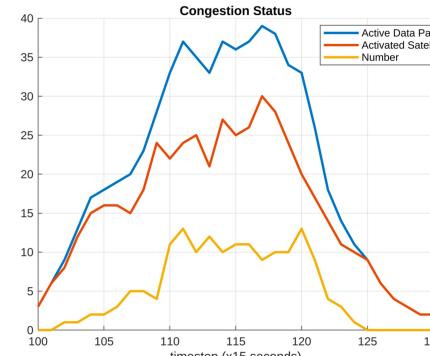
if $\max\{q'_{j1}, q'_{j2}, q'_{j3}, q'_{j4}\} = q'_{j4} \Rightarrow \pi'_j(a'_{j4}|s) = 1$

create policy_updated = $[\pi'_1 \ \ \pi'_2 \ \ \pi'_3 \ \ \pi'_4]$

if policy_updated == policy

then break \rightarrow policy_updated is equilibrium policy for all active agents

else policy = policy_updated \rightarrow return to while loop



VII. Conclusion, Discussion and Future Topics

“Connecting the Dots with the Line of Genuine Insight”



The University of Texas at Austin
Cockrell School of Engineering