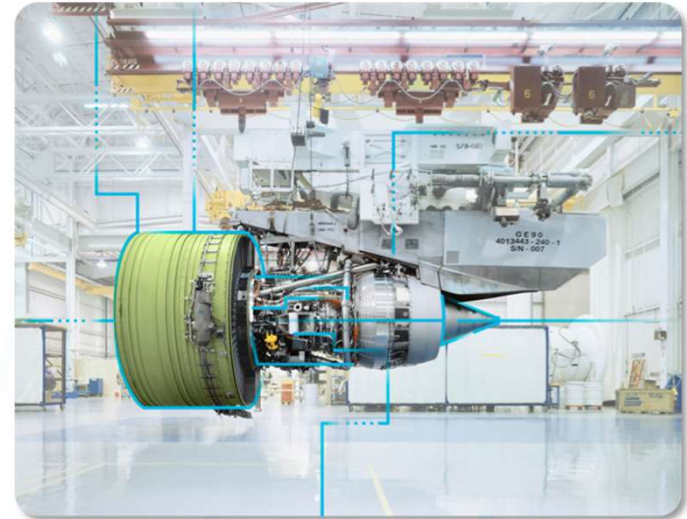




Quantum Annealing for Asset Sustainment

Annarita Giani
Aussie Schnore
GE Research



Qubits EUROPE 2019
Milan, Italy
26.03.2019

GE's technical heritage

First industrial research lab in the US

Sept. 24, 1900.

Mr. C. P. Steinmetz,
Schenectady, N. Y.

My dear Steinmetz:

I have received yours of Sept. 21 with the outline of the Plan of an electro-chemical laboratory. The proposals you make as to the investigations to be undertaken are just in the line of what I would have suggested myself.

Very truly yours,
/s/ Elihu Thomson.

EWR

GE Research today

1,000+ Scientists & Engineers

~3,000 patents /year across GE

~60,000 visitors globally/year

“I find out what the world needs, then I proceed to invent it” – Thomas Edison

Contemporary Research ...

Discovery ... world's firsts



Foreground IP

Portfolio ... differentiated tech

- ✓ **Top 10 global patents ...**
63,000+
- ✓ **Broad applicability ...**
cross disciplinary
- ✓ **Market tested ...**
product deployed

Building background IP

Prosperity ... ecosystem impact



- Generating**
1/3 of world's electricity
- Powering**
Takeoff every 2 seconds
- Curing**
16,000+ scans every minute

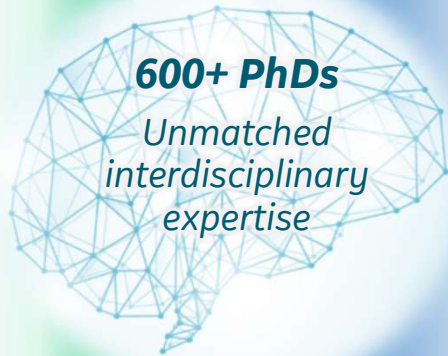
GE Businesses "+"

GE's innovation legacy & scale

GE Research ... unique value proposition

Capabilities

-  Artificial Intelligence
-  Biology & Physics
-  Controls & Optimization
-  Edge Computing
-  Electronics
-  Electric Power
-  Functional Materials
-  Mechanics & Design
-  Software & Analytics
-  Structural Materials
-  Thermosciences



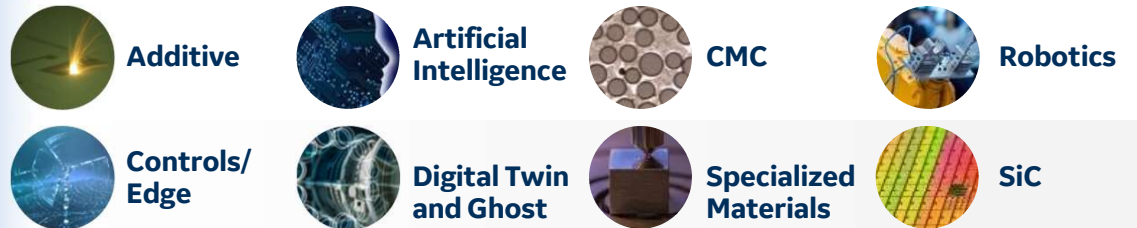
Research to Reality

Breakouts

VERTICAL



HORIZONTAL



EXPONENTIAL



Scientific Depth & Breadth ... Delivering Real Economics

GE Research

Two core research centers



A Focused, Competitive Industrial R&D Organization

Choosing the problem



The problem

Supply Chain Optimization ▶ *Logistics* ▶ *Asset Sustainment*

Many repairs are requested at the same time. Each needs a sequence of resources.

Objective Function: Minimize resource contention

Challenge: NP-hard problems, closed form solutions quickly scale in complexity beyond the ability of classical methods to provide a timely and optimal answer

Current Approach: Sophisticated algorithms together with heuristics with less than optimal results

Our Approach: Quantum Annealing and QUBO formulation of the problem to be ingested in an annealing machine to get the *optimal* set of sequences

Leveraging Previous Work on Supply Chain and Logistics

Resource Allocation Optimization

- Skilled operator
- Depot
- Warehouse
- Supplier
- Other plane
- 3D printer
- Spare parts
- Repair centers
- Bases
- Teardown centers
- Manufacturing centers

Assets

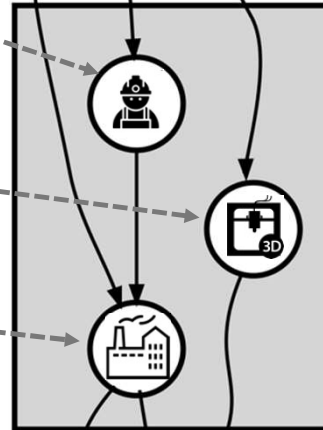


Start Repairs



Sequences
(Alternatives ways to achieve the same repair)

Resources
(Shared & unshared resources and operations)



Repairs End



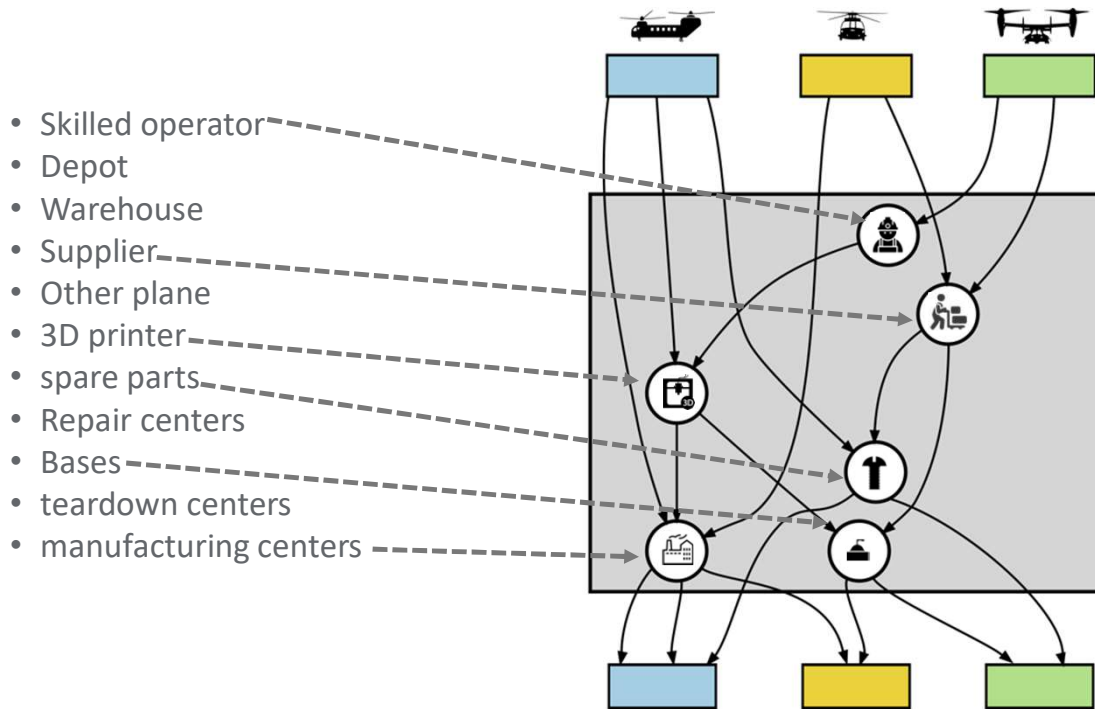
Engine needs a repair

**Find all possible solutions
(sequences of resources)**

**What is the optimal sequence
of resources that address the
repair?**

A Single Repair has Multiple Paths

Resource Allocation Optimization



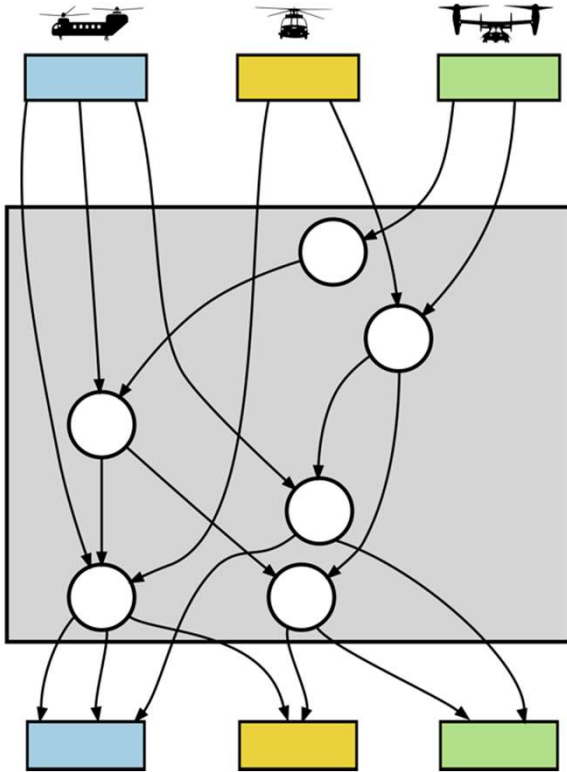
Many parts/machines to be replace/repared

Solutions are sequences of basic steps

What is the optimal sequence of resources that address all the repairs simultaneously?

Complexity Increases with Concurrent Repairs

Simple Problem



Reduce congestion at the network of repair resources

Problem size:

- **3** repairs
- **7** sequences
- **6** resources
- **128** possible choices
- **12** constrained answers

Real World Problem of Interest to GE

From Real Problem to Results

Problem Formulation

COST

$$C_{0i} = (q_{0i} + q_{1i})^2$$

$$C_{1i} = (q_{1i} + q_{2i})^2$$

$$C_{2i} = (q_{2i} + q_{3i})^2$$

$$C_{3i} = (q_{3i} + q_{4i})^2$$

$$C_{4i} = (q_{4i})^2$$

$$C_{5i} = (q_{0i} + q_{2i})^2$$

$$C_{total} = C_{0i} + C_{1i} + C_{2i} + C_{3i} + C_{4i} + C_{5i}$$

$$E_i = P_{0i} + C_{total}$$

CONSTRAINTS

$$P_{0i} = K * (q_{0i} + q_{1i} + q_{2i} - 1)^2$$

$$P_{1i} = K * (q_{1i} + q_{2i} - 1)^2$$

$$P_{2i} = K * (q_{2i} + q_{3i} - 1)^2$$

$$P_{3i} = P_{0i} + P_{1i} + P_{2i}$$

QUBO/ Ising Model

$$E(s) = \sum_i h_i q_i + \sum_{i,j} J_{i,j} q_i q_j + C$$

Upper Triangular Matrix

$$J_{i,j} = \begin{bmatrix} -9 & 18 & 18 & 0 & 0 & 0 & 0 \\ 0 & -7 & 20 & 2 & 0 & 0 & 2 \\ 0 & 0 & -7 & 2 & 0 & 2 & 0 \\ 0 & 0 & 0 & -8 & 18 & 0 & 0 \\ 0 & 0 & 0 & 0 & -7 & 2 & 2 \\ 0 & 0 & 0 & 0 & 0 & -7 & 18 \\ 0 & 0 & 0 & 0 & 0 & 0 & -6 \end{bmatrix}$$

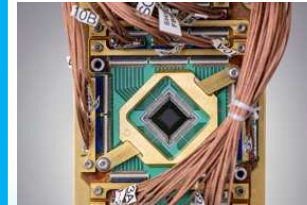
Python

```
from quboio import BinaryQuadraticModel
from quboio import BQM
from quboio import Sampler
from quboio import EmbeddingComposite

# ... (matrix definition) ...

bqm = BinaryQuadraticModel(J, C, BQM)
sampler = EmbeddingComposite(Sampler)
response = sampler.sample(bqm, num_reads=1000)
for datum in response.data("sample", "energy", "num_occurrences"):
    print(datum.sample)
    print("Energy:", datum.energy)
    print("Occurrences:", datum.num_occurrences)
```

Cloud Access

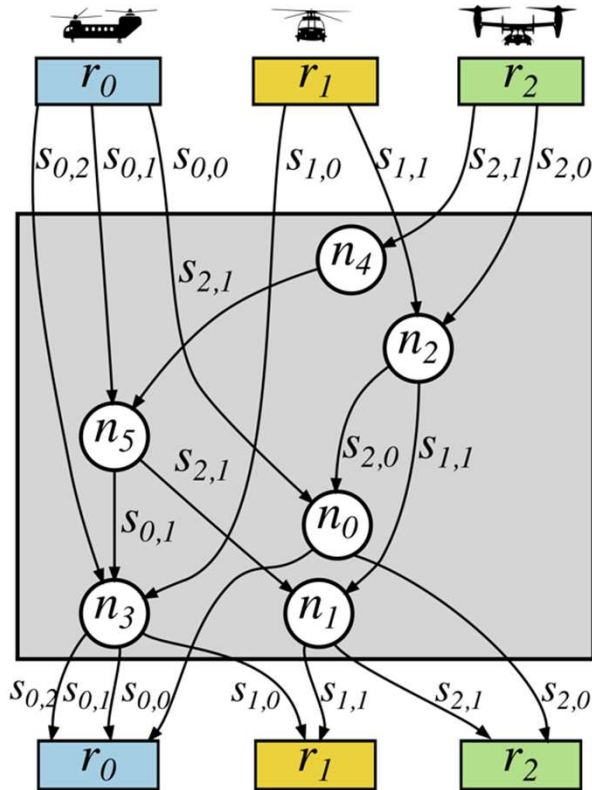


Results

Sample	Energy	Occurrences
0
1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
45
46
47
48
49
50
51
52
53
54
55
56
57
58
59
60
61
62
63
64
65
66
67
68
69
70
71
72
73
74
75
76
77
78
79
80
81
82
83
84
85
86
87
88
89
90
91
92
93
94
95
96
97
98
99

Creation of a Process for Problem Translation

Problem Formulation/QUBO/Ising Model



Qubit $q_{0,1}$ represents repair i selecting sequence of resources j

COST: Measure of Congestions at each node

CONSTRAINTS: All repairs needs to be addressed once

$$E(s) = \sum_i h_i q_i + \sum_{i,j} J_{i,j} q_i q_j + C$$

Energy = Cost + Constraints

Previous Developed Applications Helped translate this problem into QUBO

Accessing the machine and getting Results

```

1 from dimod import BinaryQuadraticModel
2 from dimod import BINARY
3 from dwave.system.samplers import DWaveSampler
4 from dwave.system.composites import EmbeddingComposite
5
6 h = {"q0_0": -9, "q0_1": -7, "q0_2": -7, "q1_0": -8, "q1_1": -7,
7     "q2_0": -7, "q2_1": -6}
8 J = {"(q0_0, 'q0_1)': 18, ('q0_0', 'q0_2'): 18,
9     ('q0_1', 'q0_2'): 20, ('q0_1', 'q1_0'): 2, ('q0_1', 'q2_1'): 2,
10    ('q0_2', 'q1_0'): 2, ('q0_2', 'q2_0'): 2,
11    ('q1_0', 'q1_1'): 18,
12    ('q1_1', 'q2_0'): 2, ('q1_1', 'q2_1'): 2,
13    ('q2_0', 'q2_1'): 18}
14 C = 27
15
16 bqmc = BinaryQuadraticModel(h, J, C, BINARY)
17
18 sampler = EmbeddingComposite(DWaveSampler())
19
20 response = sampler.sample(bqmc, num_reads = 5000)
21
22 for datum in response.data(["sample", "energy", "num_occurrences"]):
23     print datum.sample,
24     print "Energy:", datum.energy,
25     print "Occurrences:", datum.num_occurrences

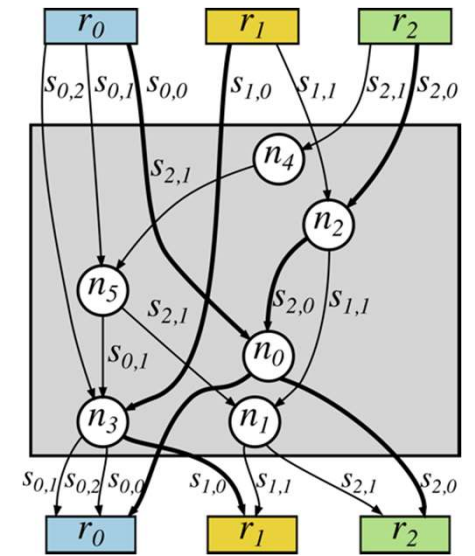
```



```

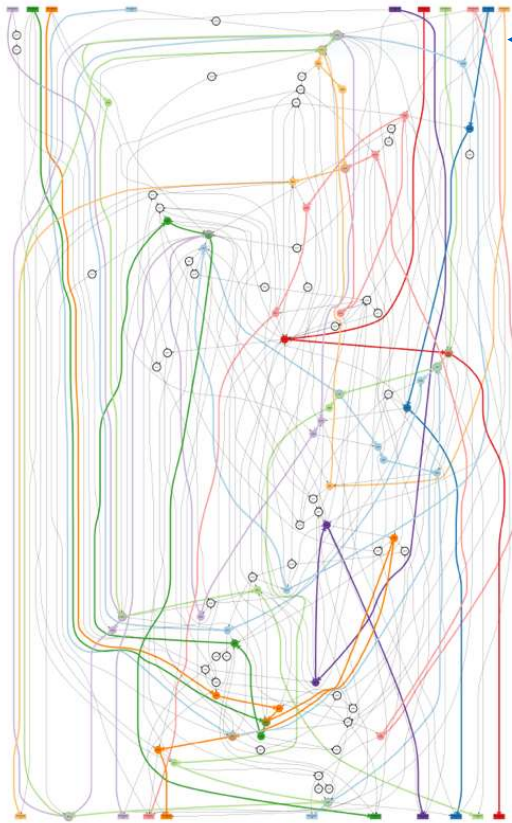
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 3 Occurrences: 18
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 3 Occurrences: 118
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 3 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 4 Occurrences: 11
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 4 Occurrences: 11
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 6 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 8 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 9 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 15 Occurrences: 2339
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 15 Occurrences: 914
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 15 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 16 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 16 Occurrences: 335
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 17 Occurrences: 178
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 17 Occurrences: 18
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 18 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 18 Occurrences: 4
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 19 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 19 Occurrences: 2
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 21 Occurrences: 3
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 21 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 23 Occurrences: 5
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 23 Occurrences: 3
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 24 Occurrences: 28
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 24 Occurrences: 3
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 24 Occurrences: 11
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 28 Occurrences: 419
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 33 Occurrences: 3
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 38 Occurrences: 1
('q0_1', 'q2_1', 'q0_2', 'q0_1', 'q0_1', 'q1_1', 'q1_1', 'q1_1') Energy: 38 Occurrences: 18

```



Getting the Optimal Choice of Resources per Repair

Scaling: Complexity Rapidly Increases

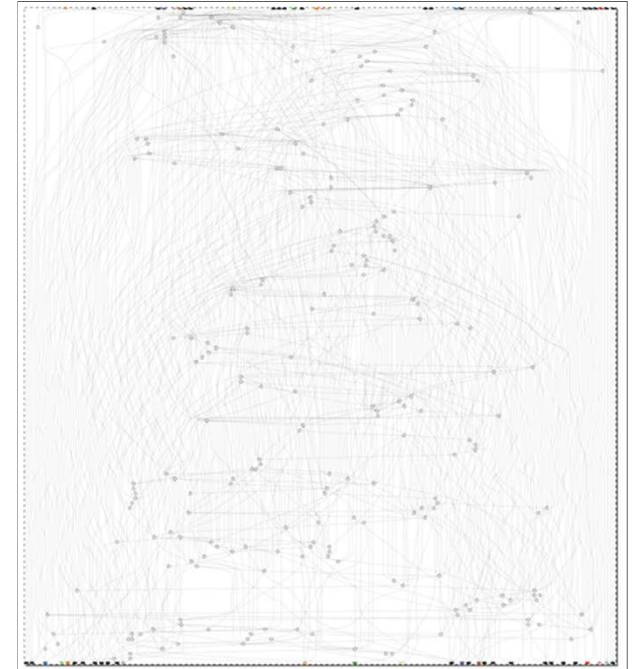


← **Problem size:**

- **10** repairs
- **28** sequences
- **94** resources
- **268,435,456** possible choices
- **12,288** constrained answers

Problem size: →

- **30** repairs
- **92** sequences
- **230** resources
- **4,951,760,157,141,521,099,596,496,896** possible choices
- **160,489,808,068,608** constrained answers



This is not even close to the size of a real sustainment problem

Classical Computing no Longer a Feasible Approach

Summary Results

Repairs	Trajectories	Resources	Combinations	Constrained Choices	Scale
3	7	6	128	12	Small Farm
10	28	94	268,435,456	12,288	Corner Garage
30	92	230	4.9×10^{27}	1.6×10^4	Dealer Service Shop

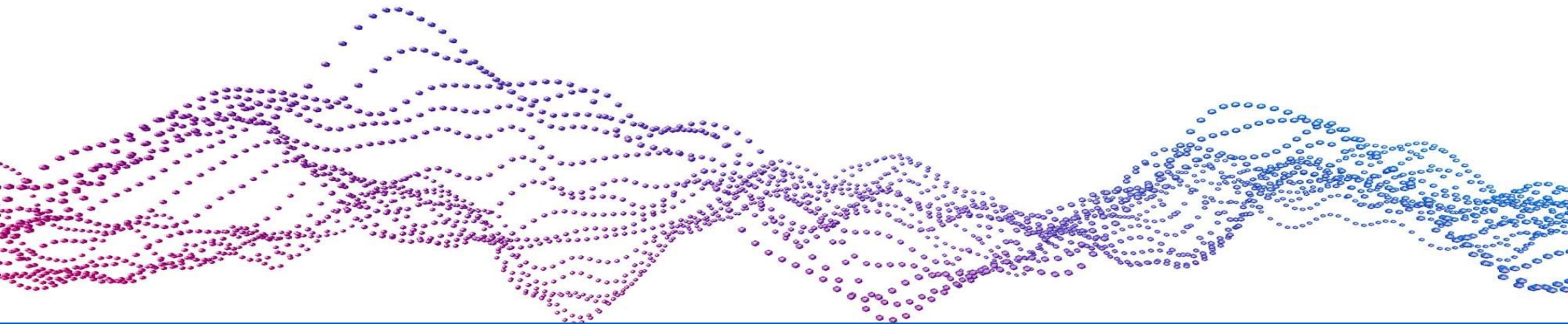
Challenges

- Deciding the problem
- Formulating the problem
- Once the problem was formulated in QUBO model
 - Value of K
 - Big Problem does not embed well into the Chimera topology
 - Chain strength

Waiting for new Topology and more Qubits

Current effort

- Making this problem more **realistic**
- New real world **applications** of interest to GE
- Real world problems on **gate model** (NISQ)





Thank you for your time

Questions?

ge.com/research/

ge.com/research/forge

Twitter: @GEResearch

