



Quantum Computing at Volkswagen: Traffic Flow Optimization using the D-Wave Quantum Annealer

D-Wave Users Group Meeting - National Harbour, MD 27.09.2017 – Dr. Gabriele Compostella

VOLKSWAGEN AKTIENGESELLSCHAFT



WE ARE

an international team working for the Volkswagen Group across the automotive value chain in the areas of Advanced Analytics and Machine Learning.

Our innovation network consists of brands' business units, experts from leading technology providers, research facilities and universities as well as upcoming Startups.



WE HIRE

creative people with skills in data analytics, AI, technology scouting, software development, DevOps, data engineering, project management, business development & agile development.





WE Python WORK Matlab WITH Scala

SPSS Modeler Neo4i SOL Solr **Tensorflow** Olik **Tableau**

WE FOCUS ON

Advanced Analytics Artificial Intelligence Machine Learning Deep Learning Data Engineering Software Development DevOps Quantum Computing Smart Enterprise NLP & Chat Bots IoT & Smart City

WE ARE HAPPY TO SPEAK TO YOU



datalab@volkswagen.de



twitter.com/VWDataLab

WE WORK

with a large amount of data in the Volkswagen Group on a state-of-the-art hybrid development environment which is highly secure, performant & flexible.

WE ARE PART OF THE VOLKSWAGEN GROUP



etc.





















VOLKSWAGEN FINANCIAL SERVICES

AKTIENGESELLSCHAFT



Team



- Dr. Christian Seidel Volkswagen Data:Lab, Munich
- Dr. Gabriele Compostella Volkswagen Data:Lab, Munich
- Dr. Florian Neukart Volkswagen Group of America Code:Lab, San Francisco
- David von Dollen Volkswagen Group of America Code:Lab, San Francisco







Team

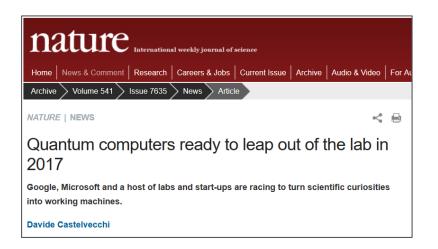


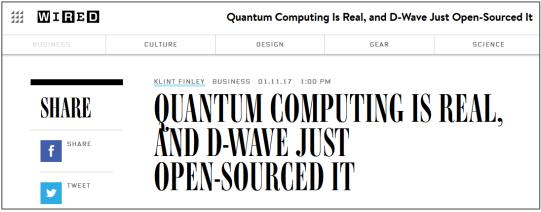




Quantum Computing – just a hype or a real thing?













The Question that drove us ...



Is there a *real-world* problem that could be addressed with a Quantum Computer?





YES: Traffic flow optimisation



Everybody knows traffic (jam) and normally nobody likes it.



Public data set: T-Drive trajectory



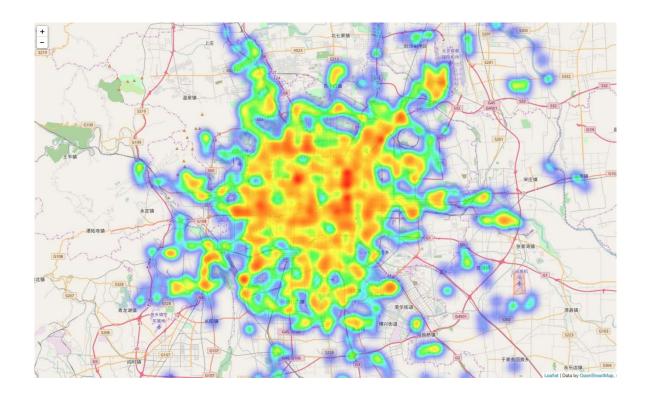
https://www.microsoft.com/en-us/research/publication/t-drive-trajectory-data-sample/

Beijing

- ~ 10.000 Taxis
- 2.2. 8.2.2008

data example:

1,2008-02-02 15:36:08,116.51172,39.92123
1,2008-02-02 15:46:08,116.51135,39.93883
1,2008-02-02 15:46:08,116.51135,39.93883
1,2008-02-02 15:56:08,116.51627,39.91034
1,2008-02-02 16:06:08,116.47186,39.91248
1,2008-02-02 16:16:08,116.47217,39.92498
1,2008-02-02 16:26:08,116.47179,39.90718
1,2008-02-02 16:36:08,116.45617,39.90531





D-Wave calculation model



Quadratic Unconstraint Binary Optimisation (QUBO)

During the quantum annealing process the system evolves to the lowest energy level. This requires the problem to be formulated as an Ising Model:

$$H_{\text{Ising}} = -\sum_{j=1}^{N} h_j \sigma_j^z + \sum_{1 \le j < k}^{N} \mathbf{J}_{jk} \sigma_j^z \sigma_k^z,$$

or as a QUBO:

$$f(x_1, ..., x_n) = -\sum_{m=1}^{N} c_m x_m + \sum_{1 \le m < n}^{N} J_{mn} x_m x_n,$$



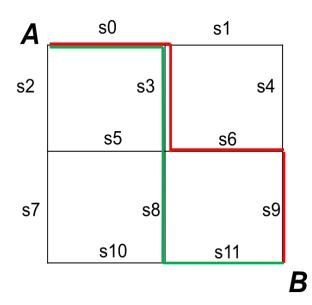
Transforming the *real world problem* for the Quantum Computer



Example:

Simplified graph structure representing a route grid.

2 cars with 3 route options on a 2 x 2 grid.



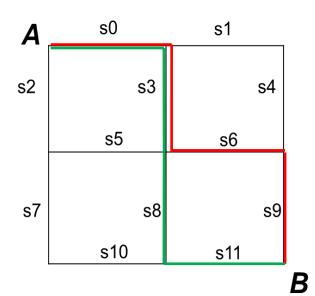
Car		Route	Binary variable
Car 1	#1:	s0,s3,s6,s9	Q ₁₁
Car 1	#2:	s0,s3,s8,s11	Q ₁₂
Car 1	#3:	s2,s7,s10,s11	Q ₁₃
Car 2	#1:	s0,s3,s6,s9	Q ₂₁
Car 2	#2:	s0,s3,s8,s11	Q_{22}
Car 2	#3:	s2,s7,s10,s11	Q_{23}



Creating the cost function for each street segment



More cars on one street lead to higher costs



Street segment	Associated cost function	Value
s0	$(Q_{11} + Q_{12} + Q_{21} + Q_{22})^2$	4
s3	$(Q_{11} + Q_{12} + Q_{21} + Q_{22})^2$	4
s6	$(Q_{11} + Q_{21})^2$	1
s9	$(Q_{11} + Q_{21})^2$	1
s8	$(Q_{12} + Q_{22})^2$	1
s11	$(Q_{12} + Q_{22} + Q_{13} + Q_{23})^2$	1
s2	$(Q_{13} + Q_{23})^2$	0
s7	$(Q_{13} + Q_{23})^2$	0
s10	$(Q_{13} + Q_{23})^2$	0

Example:

$$(Q_{11} + Q_{12} + Q_{21} + Q_{22})^2 + (Q_{11} + Q_{12} + Q_{21} + Q_{21} + Q_{22})^2 + (Q_{11} + Q_{21})^2 + (Q_{11} + Q_{21})^2$$

$$+ (Q_{12} + Q_{22})^2 + (Q_{12} + Q_{22} + Q_{13} + Q_{23})^2 + (Q_{13} + Q_{23})^2 + (Q_{13} + Q_{23})^2 + (Q_{13} + Q_{23})^2 = 12$$

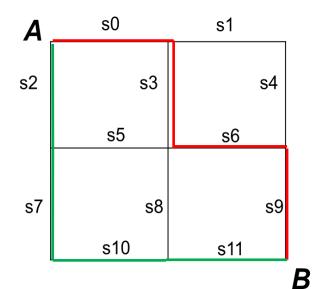
Goal: minimise the overall costs => distribute cars to different streets



Creating the cost function for each street segment



Optimal route selection:



Street segment	Associated cost function	Value
s0	$(Q_{11} + Q_{12} + Q_{21} + Q_{22})^2$	1
s3	$(Q_{11} + Q_{12} + Q_{21} + Q_{22})^2$	1
s6	$(Q_{11} + Q_{21})^2$	1
s9	$(Q_{11} + Q_{21})^2$	1
s8	$(Q_{12} + Q_{22})^2$	0
s11	$(Q_{12} + Q_{22} + Q_{13} + Q_{23})^2$	1
s2	$(Q_{13} + Q_{23})^2$	1
s7	$(Q_{13} + Q_{23})^2$	1
s10	$(Q_{13} + Q_{23})^2$	1

Minimum:

$$(Q_{11} + Q_{12} + Q_{21} + Q_{22})^2 + (Q_{11} + Q_{12} + Q_{21} + Q_{21} + Q_{22})^2 + (Q_{11} + Q_{21})^2 + (Q_{11} + Q_{21})^2$$

$$+ (Q_{12} + Q_{22})^2 + (Q_{12} + Q_{22} + Q_{13} + Q_{23})^2 + (Q_{13} + Q_{23})^2 + (Q_{13} + Q_{23})^2 + (Q_{13} + Q_{23})^2 = 8$$



Adding constraints



We want to make sure that the solution will contain only one route for each car:

 Only one binary variable per car can be 1 (otherwise the car would be on multiple routes simultaneously)

Additional constraints
K * (Q ₁₁ + Q ₁₂ + Q ₁₃ - 1) ²
$K * (Q_{21} + Q_{22} + Q_{23} - 1)^2$

- Wrong solutions must contain a penalty term
- The magnitude of the penalty term, K, is tuned based on the size of the problem (i.e. violating one constraint increases the energy of the state as if one more car was present on every road segment)

For every constraint, the overall cost function will contain a term similar to:

$$K * (Q_{11} + Q_{12} + Q_{13} - 1)^2 = K * (-Q_{11} - Q_{12} - Q_{13} + 1 + 2Q_{11}Q_{12} + 2Q_{11}Q_{13} + 2Q_{12}Q_{13})$$



Matrix formulation



Finally, the overall cost function including the constraints can be expressed in a matrix form as a function of the input binary variables:

$$F = (Q_{11} \quad Q_{12} \quad Q_{13} \quad Q_{21} \quad Q_{22} \quad Q_{23}) \begin{pmatrix} 1-K & 2+2K & 2K & 2 & 2 & 0 \\ & 1-K & 2K & 2 & 2 & 2 & 0 \\ & & -K & 0 & 0 & 0 & 0 \\ & & & 1-K & 2+2K & 2K \\ & & & & 1-K & 2K \\ & & & & & -K \end{pmatrix} \begin{pmatrix} Q_{11} \\ Q_{12} \\ Q_{21} \\ Q_{22} \\ Q_{23} \end{pmatrix}$$

This is the QUBO matrix representing the overall cost that will be minimized by the quantum computer corresponding to the 2x2 grid example.

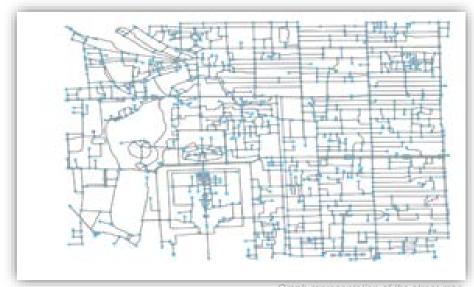


Data preprocessing



- Identification of street segments in the map by turning it into a graph using OSMnx, a Python package for street networks
- Get valid alternative routes for every car that needs to be rerouted

 Determine overlapping segments in the possible routes and define the cost function for the problem



Graph representation of the street man

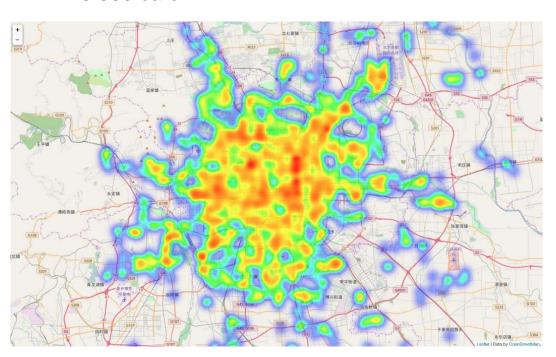


Beijing – Traffic Heatmap



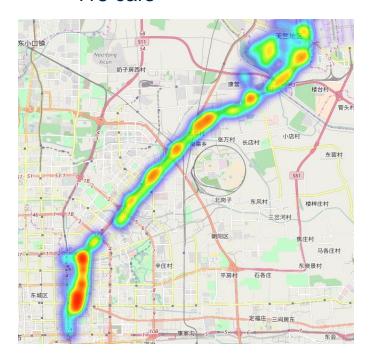
Traffic in the city

• 10.000 cars



Detail: route to the Airport

418 cars



- → We assigned each of the 418 cars 3 possible routes to reach the airport
- → Size of the problem space: 3^418



Code snippet and output



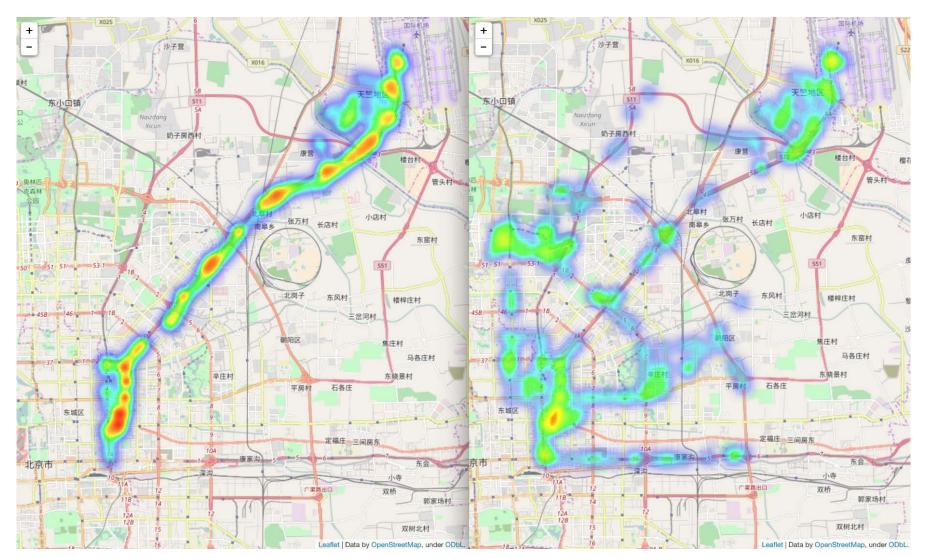
```
from dwave sapi2.core import solve_ising, solve_qubo, async_solve_ising Number of cars:
 from dwave sapi2.util import ising to qubo, qubo_to_ising, get_hardware_Number of possible routes per car:
 from dwave sapi2.embedding import find_embedding, embed_problem, unembedComplexity of traffic flow problem: 3^418
 from dwave sapi2.remote import RemoteConnection
 rc = RemoteConnection(url, token)
 solver = rc.get solver(solver name)
 # # construct Ising problem
\exists h, J, = qubo_to_ising({(i, j): qubo[i][j]}
          for i in range(nCars*nVars) for j in range(nCars*nVars)})
 ## Embedding onto hardware-like solvers:
 embedding = find embedding(J, adj)
Fif embedding is successful, continue
     if len(embedding):
      print 'Total number of physical qubits:'
       print sum(map(len, embedding))
 hemb, jemb, jchain, embedding = embed problem(h, J, embedding, adj)
 high = max(map(abs, jemb.values()))
 chain strength = 5
□for (i, j) in jchain.keys():
     jemb[(i, j)] = -high * chain strength
     params = { "answer mode": "histogram", "num reads": 1000}
     num iterations = 2
     # solve Ising problem with parameters
     answer = solve ising(solver, hemb, jemb, **params)
```

```
17-05-2017 10:33:16
                                   --- TRAFFIC FLOW OPTIMIZATION ---
                                    418
Number of constraint equations:
                                    2491
                                    1254
Number of qubits needed:
Creating quadratic unconstrained binary optimization (QUBO) problem:
Matrix size: 1254 x 1254
 [-2066 4156 4148 ...,
                                         0]
     0 -2065 4148 ...,
            0 -2070 ....
                  0 ..., -2072 4148 4148]
                  0 ...,
                             0 -2071 4150
Connecting to the D-Wave Quantum Computer in Vancouver, Canada...
D-Wave solver name:
Number of available qubits:
                               1135
Number of available couplers: 3265
Number of qubits needed (1254) > number of available qubits (1135)!
Using problem decomposition algorithm (QSage)
Starting D-Wave QPU calls
Updating route selections...
Finished successfully
```



Result: unoptimised vs optimised traffic





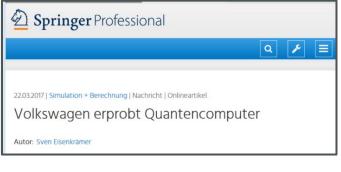


Volkswagen Quantum Computing in the news

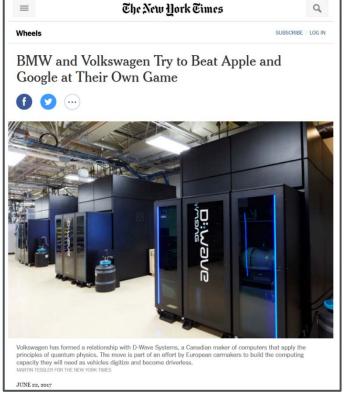














Further improvements



Due to the 2.5 months project time we treated in this test all the streets equally. This is obviously a simplification.

Additional constraints could be:

- Street capacity (highway vs alley)
- Residential zones

Dataset improvements:

- Frequent updates to react on constantly changing traffic situations (other datasets)
- More cars
- Etc.



Lessons learned



- The D-Wave Quatum Annealer can help solving real world problems (prototype was done in 2.5 months)
- transforming the real world problem into a QUBO takes the most time
- Problems larger than the chip capacity can be solved by decomposition using a hybrid solver (i.e. Qsage, qbsolv)

Due to the **chimera graph structure** of the quantum chip:

- The chip is not fully connected, so Qubit chains need to be created
- Challenging the Precision: find the right values for chain strengths (Qubit connection)
 and penalty weights

Nice to have (Christmas wishlist?):

JAVA API: would help attracting a larger audience



Publication



- Joint publication by Volkswagen and D-Wave is currently under peer review
- Draft available on the ArXiv:

https://arxiv.org/abs/1708.01625





Questions?