## Imperial College London

# Efficient Compression of Large Route Datasets using Pattern Matching

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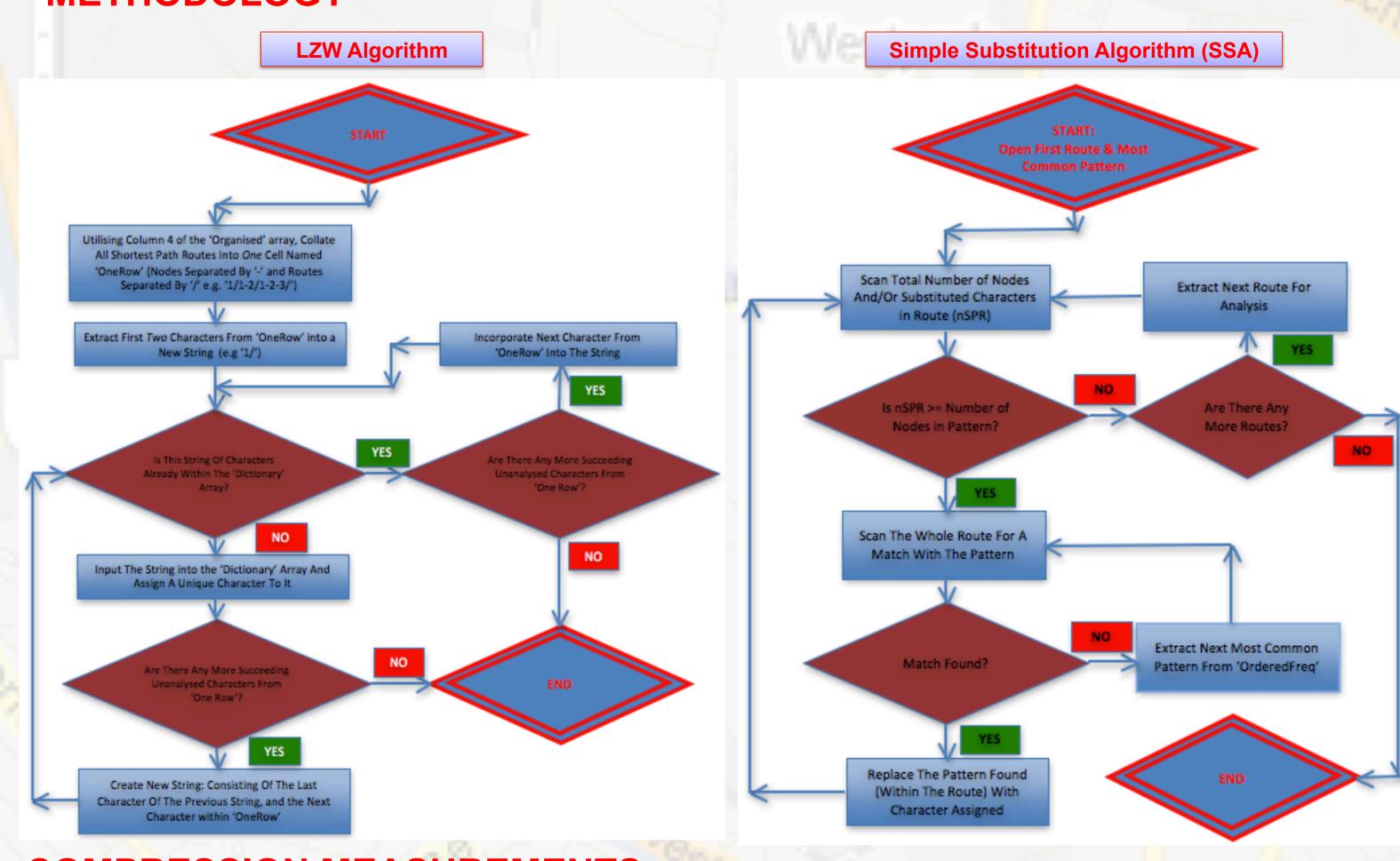
### INTRODUCTION

Transport users may require the fastest route between an origin and destination, which would involve the use of a shortest path algorithm, such as Dijkstra's. Given a sizeable number of travellers seeking routes at one time-point, 'on-demand' path generation would become infeasible, as algorithmic runtime will create delays within a queue of route requests. This research looks into ways to retrieve information to the user 'on-demand' after dataset compression techniques have been applied, given that all routes within a network are generated beforehand. These algorithms can enable a reduction in processing requirements and computer storage space within embedded/mobile devices, whilst simultaneously providing useful information quickly through decompression procedures.

#### **EXISTING COMPRESSION METHODS**

A list of compression methods were initially read over and understood, to account their suitability with this research. These included the Discrete Cosine Transform (DCT), Principal Component Analysis (PCA) and CUR Matrix Decomposition. These methods were all found not to be suitable, due to there uniqueness around image/audio compression. However, the LZW Algorithm was discovered to be highly applicable, whereby the algorithm itself 'slices' data sequentially, and references it into a dictionary.

## **METHODOLOGY**



## **COMPRESSION MEASUREMENTS**

 $CR = \frac{Number\ of\ data\ points\ in\ original\ dataset}{Number\ of\ data\ points\ in\ compressed\ dataset}$ 

 $ALTCR = \frac{Number\ of\ data\ points\ in\ compressed\ dataset}{Number\ of\ data\ points\ in\ original\ dataset}$ 

$$EDC$$
 (%) =  $100(1 - ALTCR) = 100(1 - \frac{1}{CR})$ 

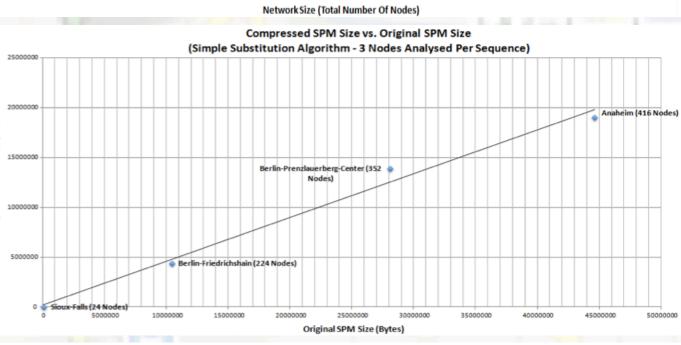
## **RESULTS**

SSA: Compressi Resul		Compression Duration (Seconds)				
<u>Network Name</u>	<u>Total</u> <u>Number of</u> <u>Nodes</u>	3 Nodes Analysed Per Sequence	4 Nodes Analysed Per Sequence	5 Nodes Analysed Per Sequence		
Sioux-Falls	24	25.046	12.838	7.718		
Berlin- Friedrichshain	224	23,565	29,281	30,587		
Berlin- Prenzlauerberg- Center	352	47,753	58,105	64,651		
Anaheim	416	217,980	530,701	750,778		

SSA: Com	pression Size I	Results	Compressed SPM Size (Bytes)			
<u>Network Name</u>	<u>Total</u> <u>Number of</u> <u>Nodes</u>	Original SPM Size (Bytes)	3 Nodes Analysed Per Sequence	<u>4 Nodes</u> <u>Analysed Per</u> <u>Sequence</u>	<u>5 Nodes</u> <u>Analysed Per</u> <u>Sequence</u>	
Sioux-Falls	24	80,232	13,824	13,824	16,128	
Berlin- Friedrichshain	224	10,453,292	4,415,488	5,318,656	4,816,896	
Berlin- Prenzlauerberg- Center	352	28,083,114	13,877,248	14,372,864	14,372,864	
Anaheim	416	44,625,714	19,036,160	20,168,648	20,057,334	

LZW Algorithm: Compression Size Results & Analysis								
Network Name	Total Number of Nodes	Original SPM Size (Bytes)	Compressed SPM Size (Bytes)	Compression Ratio (CR)	Alternative Compression Ratio (ALTCR)	Extent of Data Compression (EDC [%])	Time Take To Generate 500 Routes (Seconds)	Average Decoding Speed For 500 Routes (Seconds)
Sioux- Falls	24	80,232	2,934	27.346	0.037	96.343	28.306	89.977
Berlin- Friedrichs hain	224	10,453,292	1,690,566	6.183	0.162	83.827	45.211	372.616
Berlin- Prenzlau erberg- Center	352	28,083,114	6,199,466	4.530	0.221	77.925	49.661	503.765
Anaheim	416	44,625,714	11,580,422	3.854	0.260	74.050	51.798	645.950

350000 —				Algorithm - 3				
300000								
250000								Anaheim (416)
200000								
150000								
100000								
50000	Saur Falls (24 Nodes)			•	Berlin-Friedrichshain (2	24 Nodes)	Berlin-Pr	enzlauerberg-Center (352 Nodes)
۰.	Sioux-Falls (24 Nodes)	100	150	200	250	300	350	400 45



LZW Algorithm: Compression & Decompression Duration Results									
letwork Name	<u>Total</u> <u>Number of</u> <u>Nodes</u>	LZW Compression Duration (Seconds)	LZW Original  Decompression Duration  (Seconds)	LZW Alternate  Decompression Duration  (Seconds)					
Sioux-Falls	24	43.233	10.601	0.052					
Berlin- Friedrichshain	224	531,082.855	45,317.720	4.424					
Berlin- renzlauerberg -Center	352	1,257,058.427	139,893.669	3.997					
Anaheim	416	2,035,121.435	278,347.892	5.239					

### **DISCUSSION & CONCLUSION**

Through the results shown, it can be seen that:

- •LZW Algorithm achieves a much higher 'extent of data compression' compared to the SSA
- •LZW Algorithm takes a much longer time to complete than the SSA

The reasons for this include the nature of the LZW Algorithm: taking each small 'slice' of data and checking if it has been referenced within a dictionary. Ultimately, the LZW Algorithm achieves better compression because all sliced data has a corresponding substituted character, unlike the SSA which contains a mixture of substituted characters and 'leftover' nodes. An important note is that processing power of the computer also has an effect upon the compression duration results. Conclusively, both algorithms achieved a substantial amount of compression for the networks analysed. The implications could mean that significant compression could allow the storage of shortest path route data within embedded/mobile devices that would not need an 'online' internet connection (hence function 'offline'). The most interesting aspects of this study include how the format of data (saved within MATLAB) can drastically affect the number of bytes utilised for storage space. Information saved in 'char' format took up far less space than information saved within 'cell' or 'numeric' format.

## **ACKNOWLEDGEMENTS**

I would like to thank Dr. Angeloudis and Nils Goldbeck for their help and support throughout this project.