

Event Camera Lossless Compression for Satellite Applications

Event Data Compression on a Nvidia Jetson Orin Nano for CubeSats

Ottobrunn, 20.10.25

Antonio Junco de Haas- ge27cuy@mytum.de

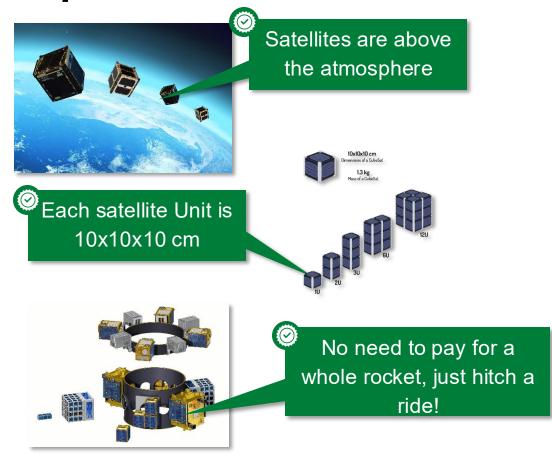




Space present unique opportunities and challenges. CubeSats allow universities to build their own space missions.

Opportunities

- Satellites can observe space without atmospheric interference.
- © Cube Satellites have brought down costs, making it possible for universities to develop satellite missions.
 - They use standardized 10x10x10cm cube Units (U).
 - Popular for universities
 - Mostly Low Earth Orbit Satellites
- Satellite launch is possible thanks to rocket ride sharing.



Satellite offer opportunities due to their unique location **in space**. Cubesats reduce costs and simplify development, allowing universities to build missions.

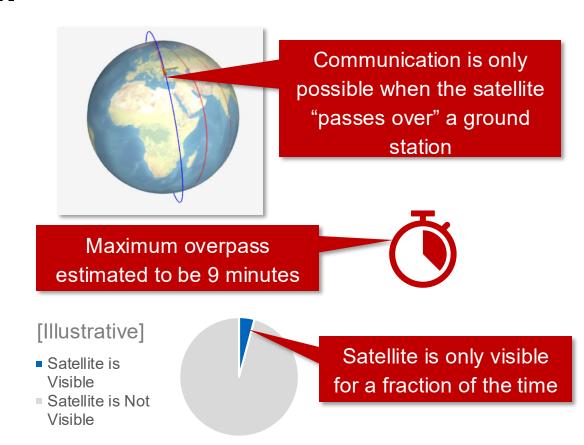




Space also presents communication challenges that must be addressed for mission to succeed.

Challenges¹

- Limited communication windows with the satellite.
- Satellite moves quickly through the sky.
- Data transmission down from the satellite (downlink) is a difficult task.
 - Satellites need to be visible to antennas on Earth to send and receive messages.



Satellite offer opportunities due to their unique location in space. However, space introduces challenges, as such data must be clear, concise and compressed.

1. For a Low Earth Orbit Satellites





EventSat is a TUM CubeSat mission that will perform space observation in a Low Earth Orbit

Mission Parameters

- University satellite mission developed by the Chair of Spacecraft Systems.
- The size of the cube satellite is 6-units (6U).
- Launch into Low Earth Orbit.
- The satellite will carry an Event Camera.
- EventSat will observe stars, planets, inter-planetary objects, other satellites and **space debris.**
 - Space debris is of particular interest as it poses risks to future satellite missions



The onboard Event Camera seeks to observe and document space observations.





EventSat will be focused on observing the surrounding space environment.

Mission Objective 1

Perform real-time detection to identify objects in the event camera field of view using computer vision.

Mission Objective 2

Develop a database of space-based event camera observations.

This objective is relevant for the thesis, as a lot of data will need to be transmitted

The onboard Event Camera seeks to observe and document space observations.





The satellite will generate huge amounts of data. Which must be downlinked to the ground station.

Regular Image



An HD picture from a regular black and white camera is 1 MB. Less than 1% of an event camera observation.

Satellite Observation



130 MB

A 44 second event camera observation can be upwards of 130 MB (after lossy compression)

Event cameras can generate huge files, as such data transmission will be a significant bottleneck for the mission.





Lossless compression will yields smaller files. Satellite can perform other operations, such as more observations.

Lossless compression can reduce data file size without deleting any information

Smaller files can be transmitted faster to the ground station

Lossy compression exists, but is not the focus of this thesis

More data can be transmitted, therefore the satellite can perform more observations.

3 Smaller transmission times means more data can be transmitted to the ground station

By utilizing lossless compression, the satellite can reduce information size without losing any data, so satellite may then prioritize other operations that can contribute to the mission objective.





An investigation for event data compression algorithms was performed, the conclusion was to create one in-house.

Algorithm Result



Proprietary compression algorithms developed **by** the event camera **manufacturers**

The only relevant event data compression algorithms found were developed by the event camera manufacturers. Leaving little demand for additional algorithm development



Not optimized for event data, as such, algorithm does not understand the data structure to the same degree as the proprietary compression algorithms

Because of the lack of third party algorithm development, and the lack of data understanding, it was concluded that an inhouse compression lossless compression algorithm would be beneficial to EventSat.





These thesis will seek to fullfill the following Resarch Contributions to deliver a successful solution

- 1
- A novel event data compression algorithm that uses a hybrid dictionary and variable-length encoding strategies. The dictionary encoding approach categorizes the data by size and compresses it in a separate section from the rest of the file. The variable-length encoding approach integrates compression instructions, indicating size right next to the data.

2

A comprehensive trade-off histogram analysis to determine the optimal compression thresholds for dictionary and variable length. The algorithm calculates file sizes for each threshold until the minimum size is found.

The Research contributions revolve around compressing the file to the smalles size possible by finding optimal data thresholds





Example Event Camera Video of the Berlin night sky.



Event Data Explanation

- Camera is panning, creating the illusion of movement
- White pixels are where brightness increased
- Blue pixels are where brightness decreased
- Stars move directly upwards
- Other space objects are moving diagonally





An Event Camera is a specialized sensor that only records brigtness changes as Events, optimal for space applications.

Example Event Camera Observation



What is an Event Camera

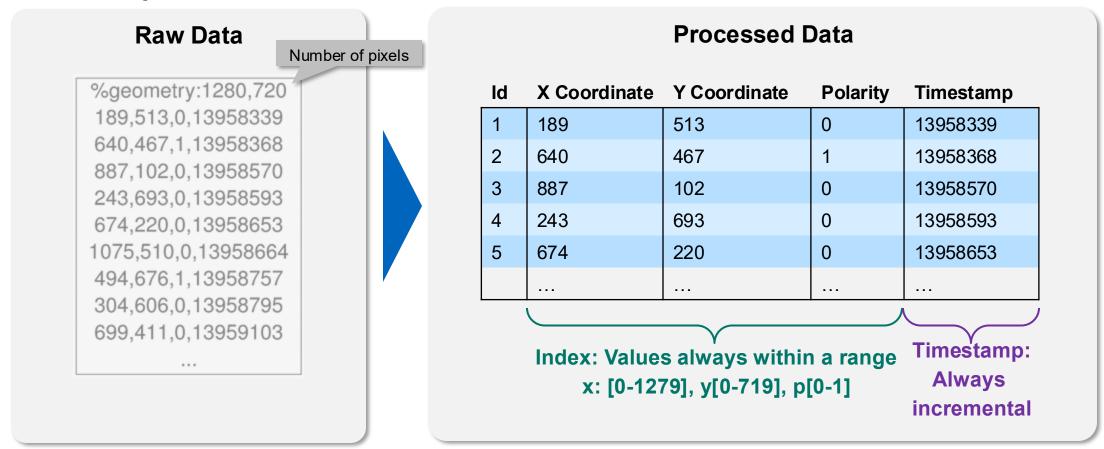
- They mimic human eyes (neuromorphic)
- They record Events, brightness intensity changes, instead of taking pictures
- They can record with ultra-low latency.
- They have the potential to surpass traditional framebased "picture" cameras by detecting fast-moving objects
- They are considered for space applications such as Space Situational Awareness (understanding the space environment around a spacecraft)

Event cameras have certain advantages for detecting objects. Potential applications is space are being considered





Events only record the pixel, polarity (brightness increased or decreased) and the timestamp where an event occurred



The geometry indicates number of pixels, **the index values** are can be any number **within the range** indicated by the geometry, **the timestamp is always incremental.**





Algorithm Overview

Events are ordered chronologically, optimal for delta encoding timestamps

Sort through the data and separate the longest events for separated compression

Write data as bits, rather than bytes

Variable-Pair **Dictionary** Delta Input Bit-Length Output **Raw Event Data Encoding Encoding Encoding** splicing **Encoding** Compressed .bin file

Receive the data from the camera

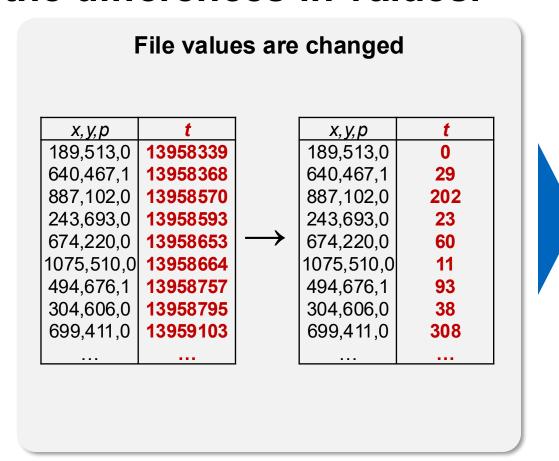
Compress the three index (x, y, polarity) numbers into a single one

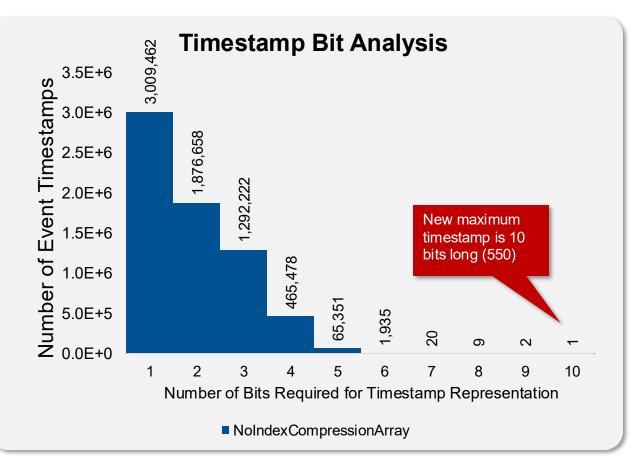
Add bit to indicate size of the remaining timestamps





Delta encoding will reduce timestamp size by only recording the differences in values.



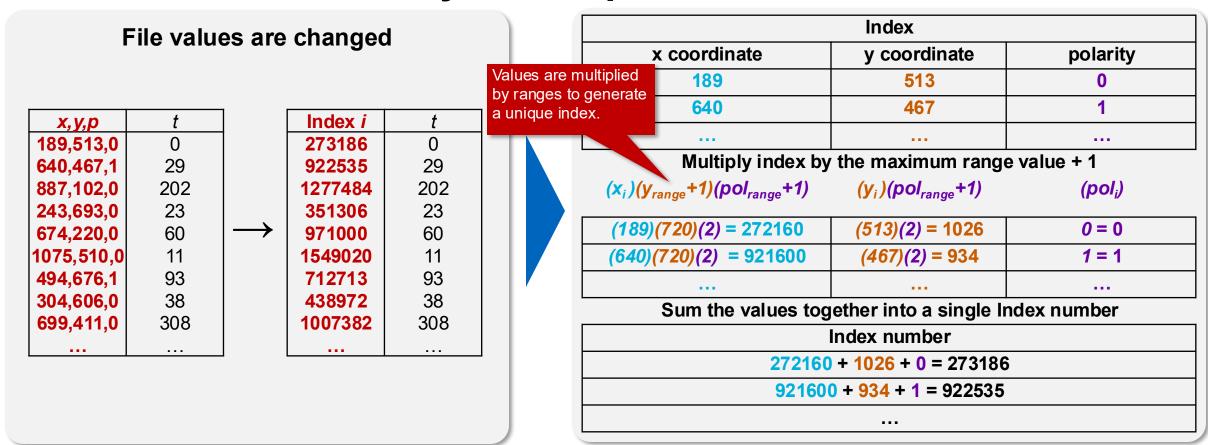


To delta encode optimally it is necessary to delve into how the data works. First data needs to be in chronological order (no negative values). Only the difference in values is recorded, if events are moved, they need a pointer to indicate location.





Pair encoding combines multiple natural numbers into a single one. It is an efficient way to compres index values.

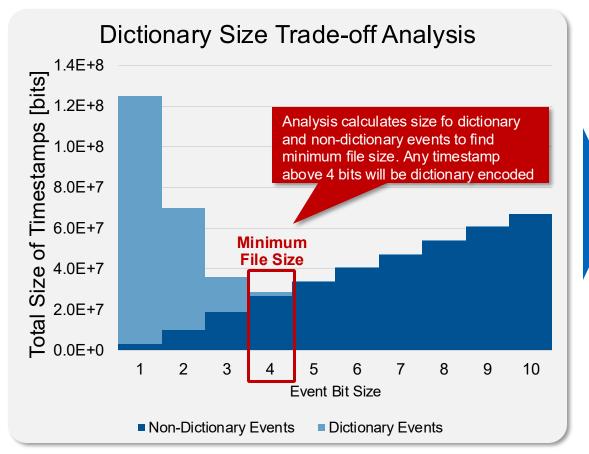


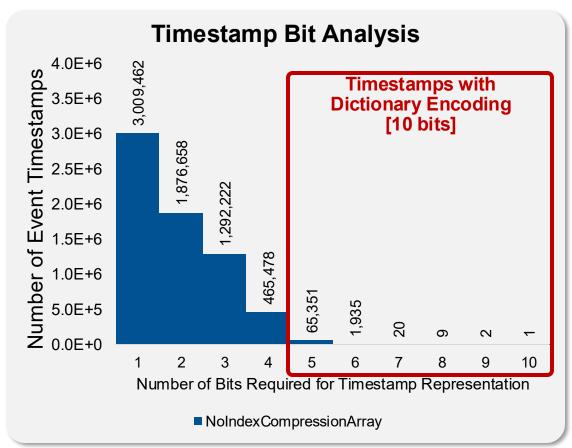
Pair encoding enables data to compress 3 integers into a 1 integer.





Dictionary encoding separates long timestamp events. Requires a threshold to indicate events to compress.



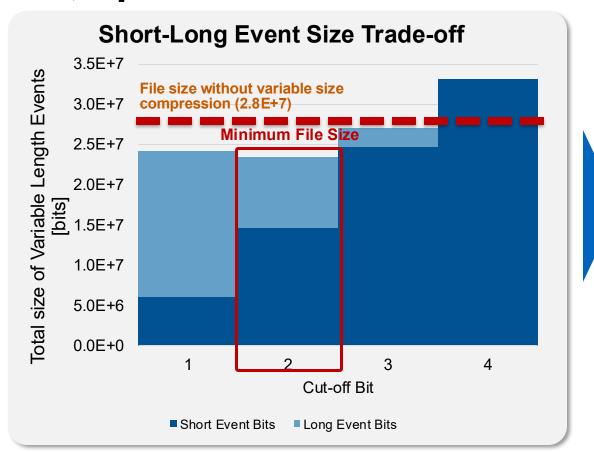


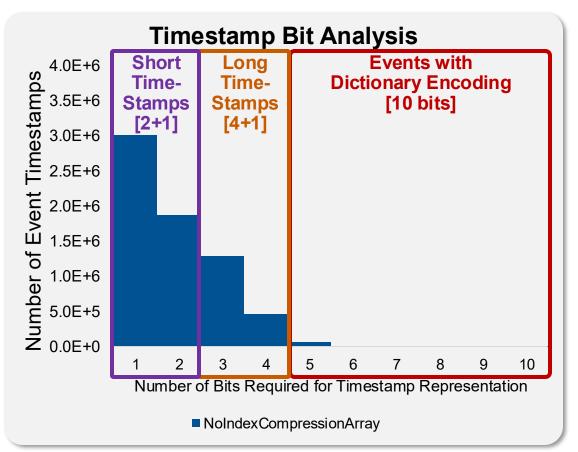
To produce the dictionary trade-off histogram, and extensive mathematical analysis was developed. File size was simulated for each bit threshold, and the optimal minimum file size is selected.





Variable length encoding adds one bit to indicate timestamp size, optimal threshold calculated.



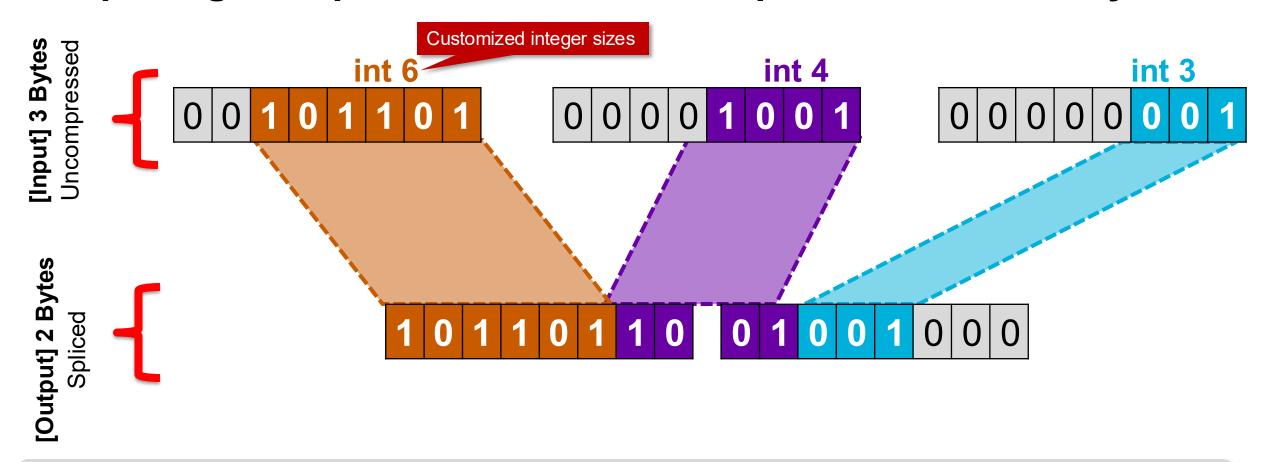


An extensive histogram analysis was also developed for the variable length, similar to the dictionary histogram but with the additional consideration that results may be larger than the original size.





Values in computers are always written in bytes (8 bits) but with bit splicing, it is possible to write multiple numbers in a byte

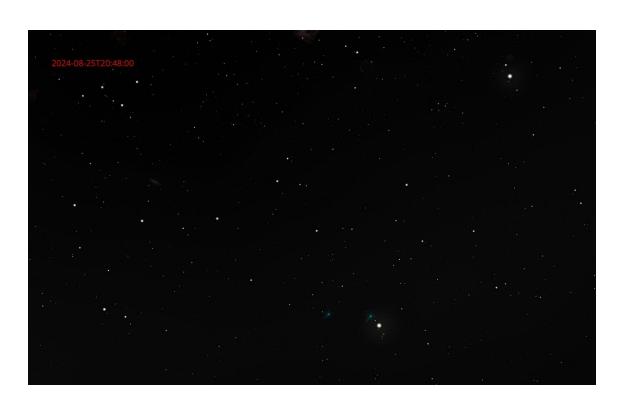


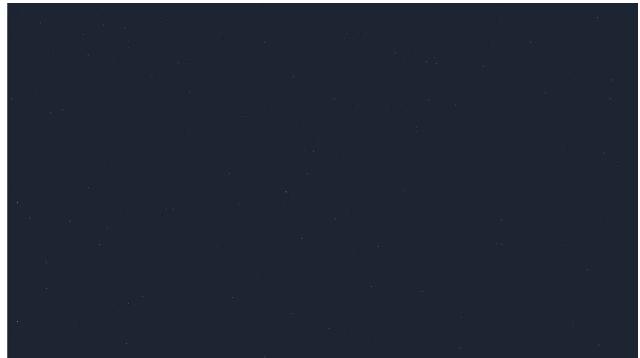
Bit splicing permits computer to write sub-byte (1 byte = 8 bits) values. This improves algorithm compression rate, especially with variable length, where short and long timestamps are both less than 8 bits.





Berlin Dataset

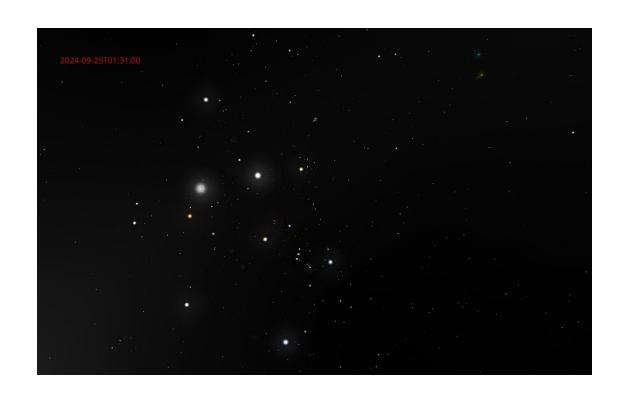








London Dataset (low visible movement)



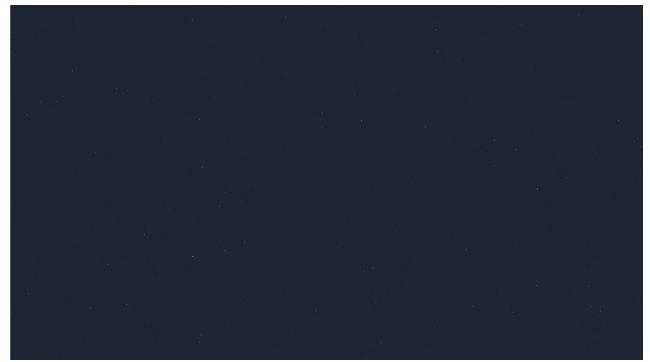






Madrid Dataset









The in-house algorithm [.bin] achieved the best compression ratio out of all the tested compression algorithms.

Algorithm	Compression Ratio [%]				
	Berlin Dataset	Spin Dataset	Madrid Dataset	London Dataset	
PROPHESEE METAVISION FOR MACHINES	70.00	75.00	4.00	4.00	
[.raw] (Camera Manufacturers)	76.29	75.60	4.96	4.96	
[.zip] (Windows OS)	68.03	67.02	64.14	64.18	
[.bin] (EventSat Team)	84.22	83.78	81.45	81.75	

The EventSat compression algorithm outperformed the competition. A higher compression ratio means better compression.





The in-house algorithm [.bin] achieved the smallest compression file out of all the tested compression algorithms

Algorithm	Compression Ratio [%]				
	Berlin Dataset	Spin Dataset	Madrid Dataset	London Dataset	
Human Readable [.csv]	133,177,219	173,610,480	8,250,704	5,121,357	
PROPHESEE METAVISION FOR MACHINES	31,577,649	42,355,357	7,841,209	4,418,273	
[.zip]	42,575,962	57,257,994	2,958,836	1,834,221	
[.bin]	21,019,730	28,167,826	1,530,245	934,837	

The EventSat compression algorithm outperformed the competition. A smaller file size means less data needs to be transmitted





To measure algorithm's performance it is important to analyze each encoding step to visualize the change in file size.

Ablation study breaks down each individual step of the algoithm

Data Format	Individual Event Sizes	Event Distribution ¹ [%]	File Size ¹ [MB]
Raw String	18, 19, 20 bytes	100	109
Event as Integers	16 bytes	100	91
Dictionary Encoding	Dictionary: 11 bytes	1	39.7
	Non-Dictionary: 7 bytes	99	
Variable Encoding	Dictionary: 11 bytes	1	39.7
	Long Events: 7 bytes	73	
	Short Events: 7 bytes	26	
UnsignedIntegers/	Dictionary: 54 bits	1	19.13
Custom Integer Length/	Long Events: 27 bits	73	
Bit Splicing	Short Events: 25 bits	26	

The combination of all the strategies in files that are approximately 20% of the original file size. This result is due to the hybrid approach between the dictionary and variable-length encoding. As well as calculating optimal thresholds.





In summary, the compression algorithm was a success with up to 84% compression rate.

Summary

- The algorithm achieved 84% compression rate. Better than the state of the art.
- The reason this result was achieved was by understanding data at a deeper level.
- I would like to detail more about the histogram analyses, as they required considerable mathematical calculations for thresholds. More details in the thesis
- Other compression strategies were tested, but were abandoned due to lack of file integrity
- Files are documented in the EventSat Gitlab

Future Work

- Dictionary compression can be improved by adding encoding instructions in timestamp, rather than utilizing an id
- Huffman encoding could reduce indexes

A lossless event compression algorithm was delivered that outperformed the competition. Resulting in files up to 84% of compression ratio





Back Up Slides

https://observablehq.com/@jake-low/satellite-ground-track-visualizer

https://www.racecar-engineering.com/articles/f1/how-to-create-a-2017-f1-car/





Indexing Experiment





Zip (what it is using)





Backup Arithmetic







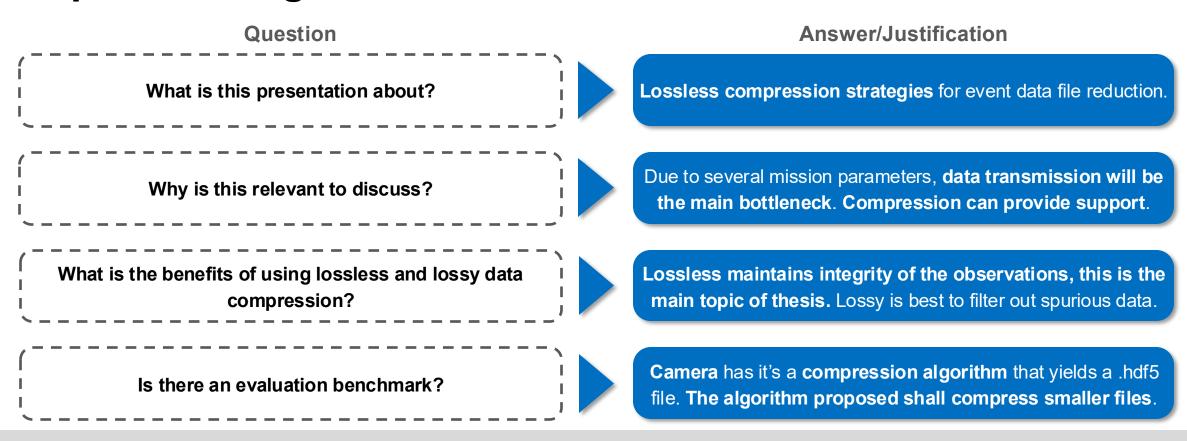


Hello





Justification and relevance for the EventSat event file compression algorithm



Compression will help mission observe longer by reducing data size for easier transmission





General information (1/2) Data bytes and the difference between signed and unsigned

Data Types and Files

- Typical numerical representation is handled in integers, long long, or short
- All variables and files in c++ (including binary files) must be written in full bytes, files are traditionally in chars

0000 0000 0000 0000 0000 0000 0000 0000

short (2 bytes): 0000 0000 0000 0000

char (1 byte): 0000 0000

Signed vs Unsigned

 The first bit of an int indicates if value is negative, unsigned integers therefore allow for double values max signed int: **0**111 1111 ... 1111 1111 = $2,147,483,648 = 2^{31}$ max unsigned int: **1**111 1111 ... 1111 1111 = $4,294,967,296 = 2^{32}$

Key considerations: Files and variables are written in bytes; Utilizing unsigned variables reduces each measurement by one bit Expected measurements number in from hundreds of thousands to millions, therefore reducing bits per measurement is necessary





General information (2/2) Bit Splicing and Encoding strategies

Custom int length and Bit Splicing

- If the recorded value of a measurement is considerably lower than integer one could apply a "customized integers length"
- Bit splicing is a technique to help write multiple values even if the data type isn't measured in bytes

Maximum recorded value: 286

int_9(using short): 1000 1111 0000 0000 [black is empty data]

Values to be recorded: 286, 160, 32

Bit spliced 1000 1111 0010 1000 0000 0100 0000

Delta Encoding

- When values are stored in ascending or descending order, delta encoding helps reduce maximum data values by storing the difference
- Works especially well when data is dense (small changes in between measurements compared to data size)

Normal Values: 13465485, 13465958, 13465974, 13465978,

13466166

Delta Encoded Values: 13465485, 473, 16, 4, 188

Delta divided by Normal Values: 1, 0.0000351, 0.0000011,

0.0000003, 0.0000140

The length of the custom int is known beforehand, and recorded somewhere when decompression is required. Delta encoding can only be used in sequential order, if values are resorted, some tracking id must be implemented





Below is the Event File Data Structure, unlike regular "RGB" cameras, events only consist of coordinates polarity and time An uncompressed event file is usually represented as a .csv or a .txt

- Data is human readable, however data filesize is large, as each written character is 1 byte (including commas and enters [¶] after each event)

	Values written in the file				
# of Event	X Coordinate	Y Coordinate	Polarity	Timestamp	Total Event Size
Event 1	272,	212,	0,	13465485¶	19 bytes = 152 bits
Event 2	1069,	703,	0,	13465958¶	20 bytes = 160 bits
Event 3	271,	455,	1,	13465974¶	19 bytes = 152 bits
Event 4	150,	76,	0,	13465978¶	18 bytes = 144 bits
Event 5	1040,	57,	0,	13466166¶	19 bytes = 152 bits

Simply transforming values to the appropriate c++ data types would be inefficient. As an integer has a total of 4 bytes (32 bits), and Boolean is 1 byte (8 bits). An event would then be 3 ints and 1 bool for a total of 13 bytes (104 bits). **Data reduction is necessary.**





Strategy 1: Timestamp Delta Encoding

- Each timestamp has the full time value written, these are also written in sequential order
- Due to this, values could be reduced by simply reading the difference

Pixel and Polarity	Timestamp		Pixel and Polarity	Timestamp
272,212,0,	13465485	_	272,212,0,	13465485
1069,703,0,	13465958	N	1069,703,0,	473
271,455,1,	13465974		271,455,1,	16
150,76,0,	13465978		150,76,0,	4
1040,57,0,	13466166		1040,57,0,	188
368,506,1,	13466205	<u> </u>	368,506,1,	39
688,123,0,	13466217	V	688,123,0,	12
989,199,0,	13466219		989,199,0,	2
478,665,0,	13466338		478,665,0,	119
462,630,0,	13466467		462,630,0,	129
80,711,0,	13466496		80,711,0,	29





Timestamp Delta Encoding

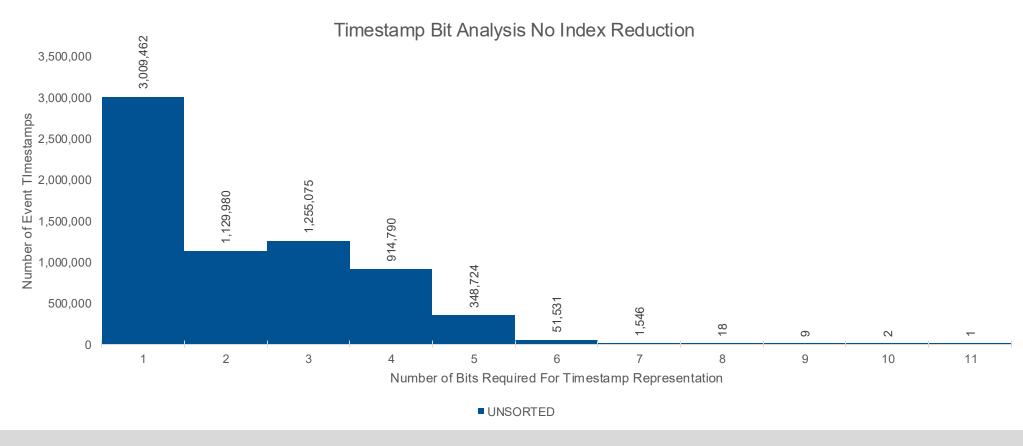
As a consequence, Timestamp is greatly reduced, in our test the largest value (without counting the initial)
was of 1440, which can be represented with 11 bits rather than the largest uncompressed timestamp
of 28237230, which needs 25 bits.

Pixel and Polarity	Timestamp	_	Pixel and Polarity	Timestamp
272,212,0,	13465485	-	272,212,0,	13465485
1069,703,0,	13465958	N.	1069,703,0,	473
271,455,1,	13465974	/	271,455,1,	16
150,76,0,	13465978		150,76,0,	4
1040,57,0,	13466166		1040,57,0,	188
368,506,1,	13466205	<u> </u>	368,506,1,	39
688,123,0,	13466217	V	688,123,0,	12
989,199,0,	13466219		989,199,0,	2
478,665,0,	13466338		478,665,0,	119
462,630,0,	13466467		462,630,0,	129
80,711,0,	13466496		80,711,0,	29





After Delta encoding a file, the number of bits required to represent an event will be reduced



Key Takeaway: After delta encoding, the maximum number of bits required is only 11





Visual Representation of Pair Enoding

Quick example, imagine we have 3 variables **x**, **y**, **z** that each can have a three values: **0**, **1**, **2**

As such one possible representation of x, y, z would be a 6 bit number: 100110_2 in binary

But with arithmetic encoding it can be reduced to a **5 bit number: 10111**2 in binary

To decompress simply use the modulo function and floor division

Multiply original values by maximums + 1

$$x = 2$$

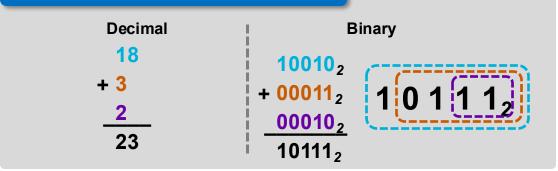
 $y = 1$
 $z = 2$

$$(2)(y_{max}+1)(z_{max}+1) = (2)(3)(3) = 18$$

$$(1)(z_{max}+1) = (1)(3) = 3$$

$$(2) = 2$$

Sum the values together



Decompress by using modulo (%) and floor division ([/])

$$z = (23 \% (z_{max}+1)) = (23 \% 3) = 2 | [23/3] = 7$$

 $y = (7\% (y_{max}+1)) = (7 \% 3) = 1 | [7/3] = 2$
 $x = 2 = 2$



Visual Representation of Pair Enoding

Index		
x coordinate	y coordinate	polarity
189	513	0
640	467	1
		•••

Multiply index by the maximum range value + 1

$(x_i)(y_{range}+1)(pol_{range}+1)$	$(y_i)(pol_{range}+1)$	(pol _i)
(189)(720)(2) = 272160	(513)(2) = 1026	0 = 0
(640)(720)(2) = 921600	<i>(467)(2)</i> = 934	1 = 1
•••		

Sum the values together into a single Index number

Index number		
272160 + 1026 + 0 = 273186		
921600 + 934 + 1 = 922535		
•••		





Arithmetic Encoding for Index Pixel and Polarity compression

- Pixels and polarity can be represented as a single number (X × 720x2) + (Y × 2) + (POL × 1).
- This is a "Customized variable base" where Polarity is [0,1], Y [0,719], X [0,1279].
- Therefore pixels can be represented by a unique number from [0,1843200] in 21 bits

Pixel and Polarity	Timestamp		Pixel and Polarity	Timestamp
272,212,0,	13465485		392104	13465485
1069,703,0,	473	N	1540766	473
271,455,1,	16		391151	16
150,76,0,	4		216152	4
1040,57,0,	188		1497714	188
368,506,1,	39	<u> </u>	530933	39
688,123,0,	12	V	990966	12
989,199,0,	2		1424558	2
478,665,0,	119		689650	119
462,630,0,	129		666540	129
80,711,0,	29		116622	29





Run Length Encoding for Polarity

It is possible to further reduce the Index by utilizing Run Lenth Encoding. First separate the values [0,1] and record how many events are on each. Similar to how one orders words alphabetically

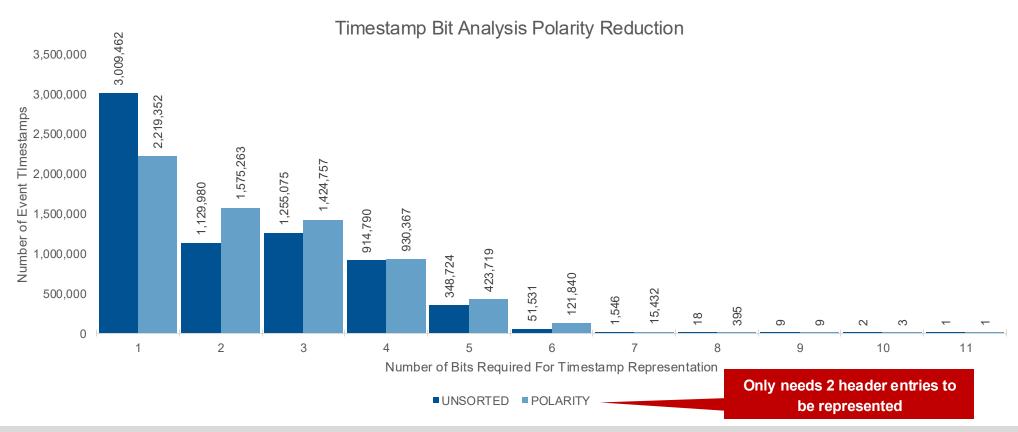
Sort the events by the Index Polar	ity			' 0, 3584555 RITY 1	
Pixel and Polarity	Timestamp		F		
272,212, 0 ,	13465485	ity 0:	272,212, 0 ,	13465485	
1069,703, 0 ,	13465958	Polarity 0: 5,579	1069,703, <mark>0</mark> ,	13465958	After sorting by
271,455, <mark>1</mark> ,	13465974	er for 3,355	150,76, 0 ,	13465978	polarity, timestamp is
150,76, 0 ,	13465978	Header for F 3,355,	1040,57, 0 ,	<mark>13466166</mark>	no longer sequential
1040,57, 0 ,	13466166	=			
368,506, 1 ,	13466205	arity 1	271,455, <mark>1</mark> ,	13465974	
688,123, <mark>0</mark> ,	13466217	Header for Polarity 3,355,559	368,506, 1 ,	13466205	
989,199, 0 ,	13466219	der fo 3,35	41,65, 1 ,	13467692	
		Неас	1019,256, <mark>1</mark> ,	13467785	

This will reduce each event entry by 1 bit as long as there is a header that indicates how many polarity 0 an 1 events exist





Because Run-Length Encoding and Delta encoding both require sorting, they reduce the effectivity of each other

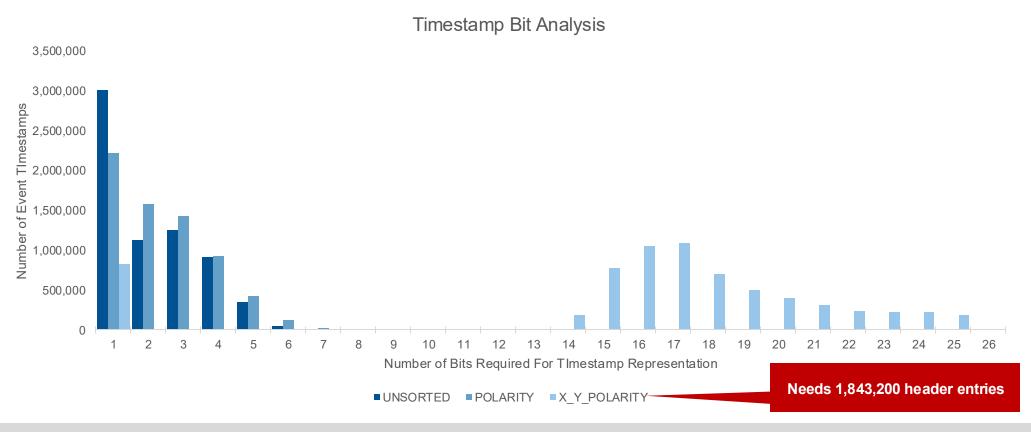


Unsorted values require less bits for timestamp representation, while polarity sorted events require more. However, in this particular example they both need a maximum of 11 bits to represent the timestamp





To show the incompatibility, a Run-Length encoding for XYPol will fully eliminate the Index, but larger Timestamps remain

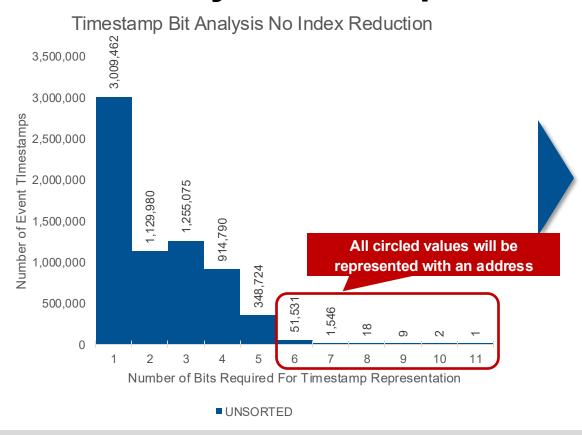


If the algorithm were to **fully remove the index**, **timestamps** would **require 26 bits** to be represented **plus an additional header** to record the number of events for each pixel and polarity must be implemented, **a balance between index**, **timestamp**, **and header must be calculated**





Dictionary Encoding can be utilized to reduce the outlier values, as they can be represented with another number



- The last 4 bit buckets in the graph are sparse, and as such many values are completely empty.
- Dictionary Encoding can allow us to represent these large values with a smaller "pointer" addresses if we utilize an id system

Event Id	Timestamp
400,578	290
1,752,005	514
2,278,656	3977

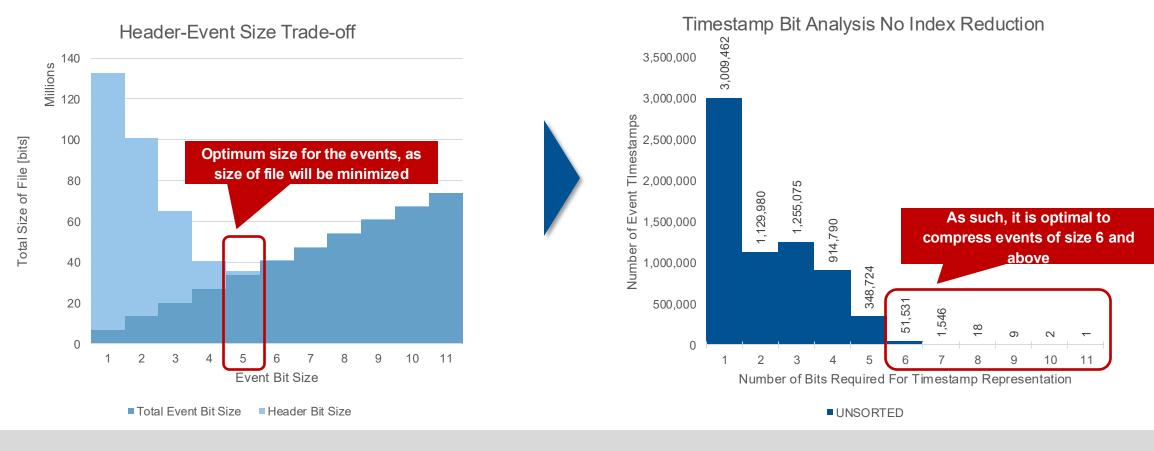
- For this encoding to work data must be sparse, as each pointer will need to represent a large number for the Event Id (size will be max number of events)
- There are some cases where it could be encoded into 7 bits, if enough zero values are available, but a further analysis is required to show the frequency of each number
- Will be impacted by Run-Length encoding

Afterwards the maximum bit necessary for timestamps will be 8 instead of 11





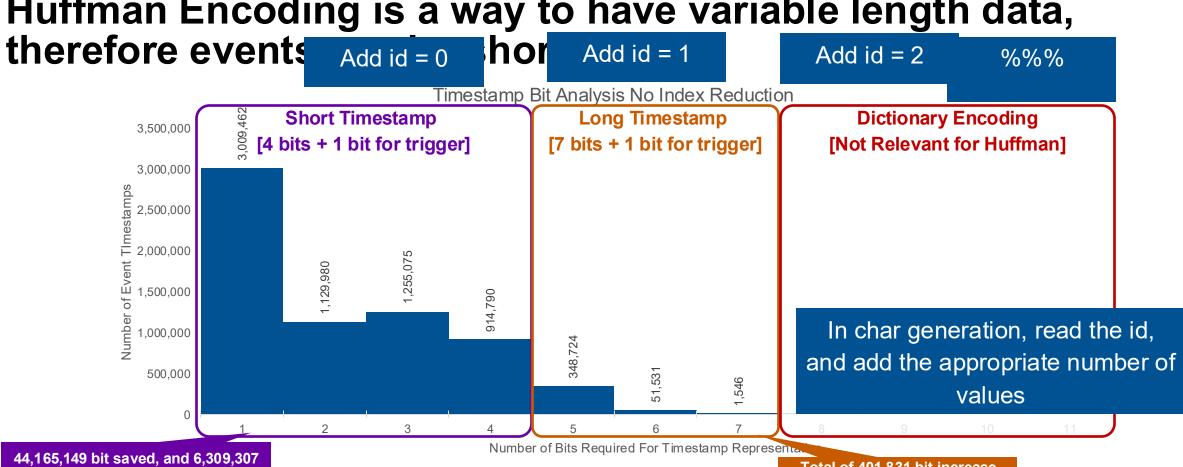
To find the optimal event size a histogram analysis is calculated with the sum of the header and event entries







Huffman Encoding is a way to have variable length data,



bit increase (due to trigger)

UNSORTED

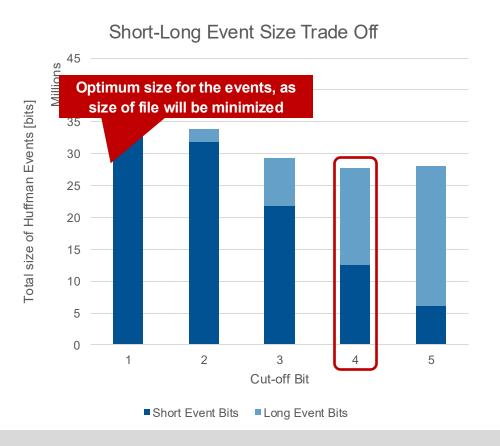
Total of 401,831 bit increase (due to trigger)

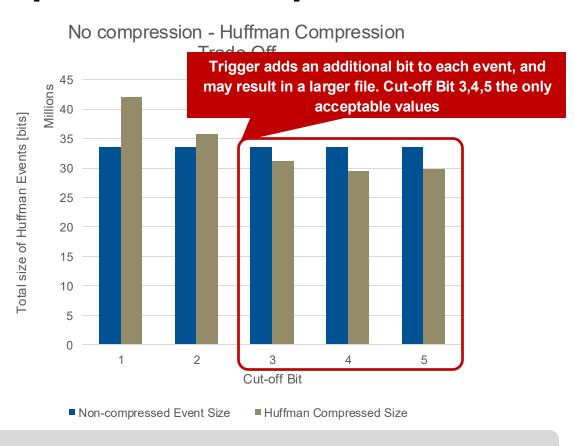
The Huffman encoding will add a trigger [0,1] that will indicate if an event has a long or short timestamp. The length of the short/long timestamp will be file specific. The choice of the trigger locations will be influenced by the Dictionary Encoding





To find the optimal short-long event size a histogram analysis must be made, as well as a non-compression comparison





Afterwards the maximum bit necessary for timestamps will be 8 instead of 11





These encoding strategies will be utilized with different parameters to find an acceptable filesize for downlink

Huffman, and Dictionary are still a work in progress





Should it be a bit every single entry? Or should it be a trigger flag that executes after a certain amount of entries?



Future strategies to be implemented (1/2) Data reduction by adding indexes

Dictionary based encoding

- Dictionaries work by having an index that represents values.
- This can compress long and complex that are repeated often into smaller ones.
- Not all values need to have an index

The most common word in English is **The**, a dictionary encoding would assign it a number **[1]** to compress a sentence. When **1** stars shine over **1** lake, **1** stargazer went to observe **1** celestial objects, and record **1** movements in **1** journal.

Variable Length Data

 Most measurements don't need the full memory allotted to them, having some kind of indicator for a reduced memory entry to minimize empty space

0 index means only assign 4 bits per entry measurement 1 index means assign max 8 bit per entry measurment

Entry 1: 0 1101 0001 1000 | Total 13 bits

Entry 2: 1 0010 1000 0001 0011 0001 1111 | Total 25 bits

Entry 3: 0 1000 0011 1010 | Total 13 bits

These strategies are **heavily dependent on the specific details of the data**, and will **yield different results** even if the instrument and measurement time doesn't vary





Future strategies to be implemented (2/2) Iterating to find the best compression strategy for file

(C) Iterative File Specific Encoding

Encode a file with different strategies multiple times to see which ones yield a smaller result file

Some strategies such as variable length data benefit from being run multiple times (iterative) to find optimal result.



TBD

TBD

Running the compression algorithm multiple times will **utilize more energy**, **however** this is not the current bottleneck projected for the mission, as such **data downlink budget takes priority**





Agenda - Content

- Esda
- Sadas
- Sadasdf
- Sa
- As
- sa





Status Bus

Summary of subsystem (EPS..working, first powered, not purchased)





Status Payload

- We worked a lot
- What we learned with results:
 - What we can observe
 - Filtering is not enough → Star tracking
 - Lara does the slides





Status Payload





Challenge - Mission Objective

- FIRST: We are not aligned in the objective mission
- Vincenzo presents





Challenge - Time

THEN: Time challenge Jaspar does the slide Vincenzo presents





Solution Propose

- Voluntary students: Reward for them?
- Thesis students: Can we do thesis based on EventSat that do not have scientific output (not Master's Thesis) Semester Thesis needs to have a scientific output, practical research course/teamwork don't need scientific output?
- Jaspar does the slide

