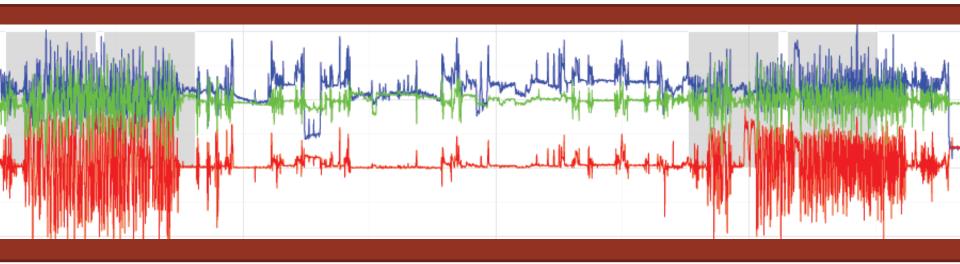
Information extraction and transparency in big data processing

Developing new endpoints for mobility is an important strategic aim for many groups both in industry and academia and the focus of a growing field. Bringing those novel endpoints to health authority acceptance for clinical decision making will require a concerted effort from this research community. This in turn will require openness and transparency; sharing data, methods and findings. Here we discuss challenges within the field to such an open approach and give examples of how they might be overcome.

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Information extraction and transparency in big data processing

- 20150305
- Ieuan Clay on behalf of the "ActiTeam"



What do we mean by transparency?

- Transparency aims to make the process of research clearly visible to all readers.
 - including data collection, coding, and analysis
- Important for:
 - Peer-review publication process
 - Collaboration
 - Regulatory submission process

...any situation where reproducibility is important...

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Why transparency? Perspective from Clinic/Industry

- Regulatory trail for the FDA
 - Reproduce any aspect of evidence supporting a submission
- Contributing to a growing field
 - Crowd sourcing
- Collaboration with TRIUM

Cross divisional, evolving team







Challenges to transparency?

http://ejournals.bc.edu/ojs/index.php/jtla/pages/view/transparency

- Protection of human subjects (e.g., extent to which data can be made available without infringing on the rights of subjects) – can we even anonymize accelerometry?
- Providing access to analytic tools (e.g., Matlab, SPSS, etc.) that are copyrighted – should we even use them?
- Providing access to tests and testing tools that are copyrighted or under development – how far can we standardise QC in a diverse and evolving field?
- Providing access to large data sets (e.g., TIMSS or NAEP) used as part of a research study – should we develop open datasets, like kaggle (<u>http://mayer.pro/t-SNE-Samsung</u>)?
- Developing community spirit and direction where is the consensus/synergy in such a diverse and developing field?
 - ...the potential uses of accelerometry are so huge, expert guidance and direction is crucial (clear pull/push)
 - ...this opens the field to many contributors



Transparent tools

- Open) Programmatic frameworks
- Documentation frameworks and standards
- Version control and other good practice
- Data sharing platforms and formats
- Ontologies
- Good project management, e.g. Role transparency within diverse teams



Contents

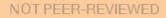
- Intro
 - Project
 - Technology & TRIUM collaboration
 - Team
- Focus on Analytical Challenges
 - Creating an environment for open collaboration
 - Information extraction from clinical tests



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Physical activity

Peer Preprints

WHO Fact sheet N°385 February 2014

- Physical activity is defined as any bodily movement produced by skeletal muscles that requires energy expenditure.
 - Including activities undertaken while working, playing, carrying out household chores, travelling, and engaging in recreational pursuits.
 - At all ages, the benefits of being physically active outweigh potential harm, for example through accidents.

Some physical activity is better than doing none.



Doorlow	2010	
Peer Preprints	Disorder NOT PEER-REVIEWED	Mean rank
Physical inactivity		(95% UI)
	1 Low back pain	1·1 (1 to 2)
	2 Major depressive disorder	1.9 (1 to 3)
	3 Iron-deficiency anaemia	3-3 (2 to 6)
- I ne tourth leading risk factor for global -	4 Neck pain	4·3 (3 to 7)
	- 5 COPD	5-8 (3 to 10)
	6 Other musculoskeletal disorders	5·9 (4 to 8)
 Appr. 3.2 million people die each year due to inactivity 	7 Anxiety disorders	6-4 (4 to 9)
	8 Migraine	8-9 (6 to 15)
	- 9 Diabetes	9·1 (6 to 13)
Physical inactivity is on the rise in many countries.	10 Falls	10-1 (7 to 14)
	11 Osteoarthritis	12·3 (9 to 17)
	12 Drug use disorders	12.5 (9 to 16)
20% to 30% increased risk of death compared to people who engage in at least 30 minutes of moderate intensity physical activity on most days of the week.	13 Hearing loss	13·5 (7 to 20)
	14Asthma	15·3 (10 to 20)
	15 Alcohol use disorders	15·8 (12 to 21)
	16 Schizophrenia	16-0 (9 to 22)
	17 Road injury	16-1 (12 to 20)
	18 Bipolar disorder	16-6 (9 to 23)
Physical inactivity is the main cause for approximately:	19 Dysthymia	18-6 (13 to 26)
	20 Epilepsy	21.8 (18 to 27)
	21 Ischaemic heart disease	21.9 (17 to 29)
 21–25% of breast and colon cancers 	22 Eczema	22-3 (16 to 35)
 27% of diabetes 	23 Diarrhoea	23·1 (19 to 28)
 30% of ischaemic heart disease 	- 24 Alzheimer's disease	25·9 (21 to 33)
	- 25 BPH	26-3 (20 to 35)
	26 Tuberculosis	Communicat
Vos et al. Lancet 2012; 380: 2163–96	27 Neonatal encephalopathy*	neonatal, and
8 HMI: Acceleramics meets Genomics Lleuan Clay J 20150305 L Information Extraction	n and Transparaney Business Ise	Non-commu

8 | HMI: Acceleromics meets Genomics | Ieuan Clay | 20150305 | Information Extraction and Transparency | Business Use Only PeerJ Preprints | https://doi.org/10.7287/peerj.preprints.2546v1 | CC BY 4.0 Open Access | rec: 21 Oct 2016, publ: 21 Oct 2016 Communicable, maternal, neonatal, and nutritional disorders

Non-communicable diseases

% change (95% UI)

43 (34 to 53) 37 (25 to 50) -1 (-3 to 2) 41 (28 to 55)

46 (32 to 62) 45 (38 to 51) 37 (25 to 50)

40 (31 to 51)

68 (56 to 81)

46 (30 to 64)

64 (50 to 79)

40 (27 to 54)

29 (22 to 36)

28 (21 to 34)

32 (16 to 50)

48 (37 to 60)

30 (13 to 49)

41 (31 to 51)

41 (34 to 48)

36 (27 to 47)

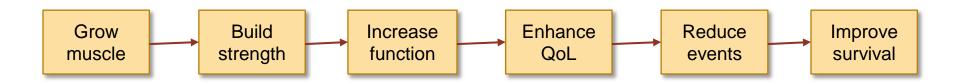
48 (40 to 57) 29 (19 to 39)

5 (-1 to 11) 80 (71 to 88)

84 (48 to 120)

Injuries

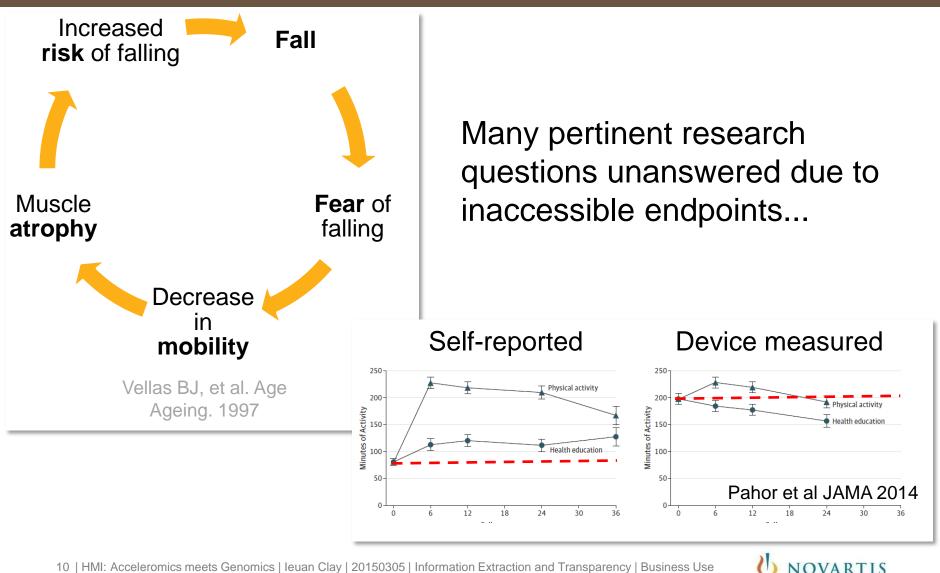
Challenges for study design



Establishing the chain of evidence



Falls and deterioration of physical activity

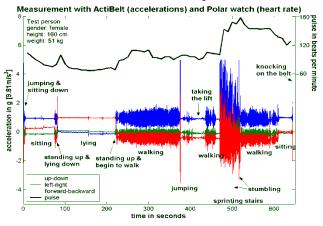


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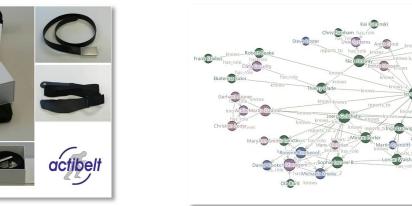
The challenge: Added value of accelerometry

Objectively captures real-world movement patterns throughout the day

Accelerometry



- Define a new field of research for drug development
- Establish new, robust clinical endpoints together with the scientific community
- Set new registerable standards for phase III trials









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External collaboration ActiBelt and TRIUM



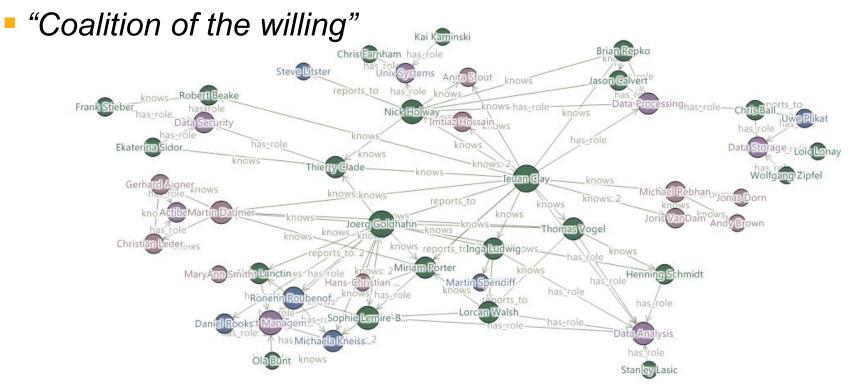






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Internal working group ActiTeam



- Jörg Goldhahn, Daniel Rooks, Sophie Lemire-Brachat, Valerie Lanctin, Miriam Porter, Thomas Vogel, Henning Schmidt, Inga Ludwig, Lorcan Walsh, Thierry Clade, Nick Holway, Ieuan Clay
- Collaborators at TRIUM / The Human Motion Institue
- Collaborators across NIBR

Peer Prepri

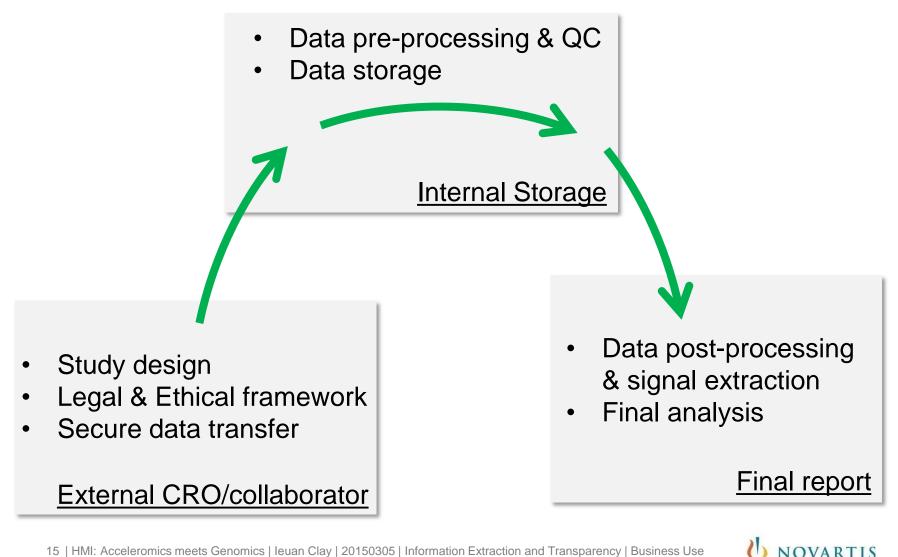
Challenges

- FDA/regulatory approach
- Translational strategic approach
- Collaborative challenges
- Competence challenges
 - Technical:
 - Data transfer & volumes
 - Data storage/access/security
 - Analytic challenges:
 - Data QC
 - real world streaming data
 - Variability & complexity
 - etc



Data Flow: overview

Peer Preprints



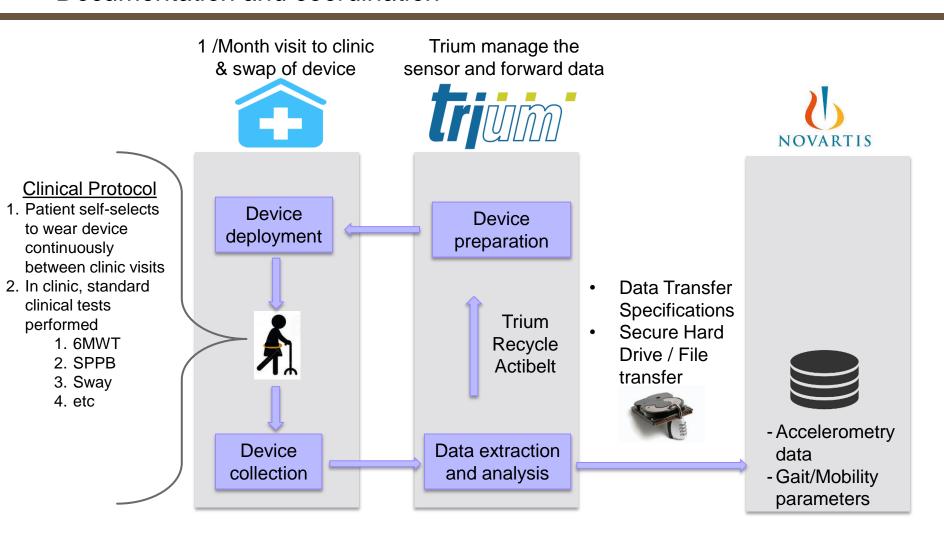
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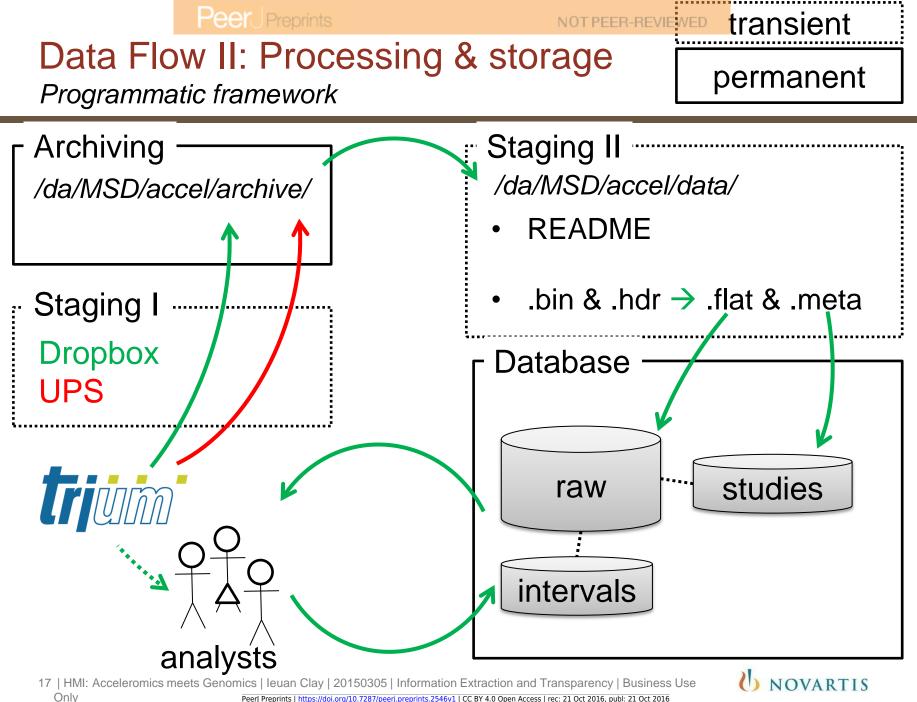
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Data Flow I: External to Novartis

Lorcan Walsh & Miriam Porter

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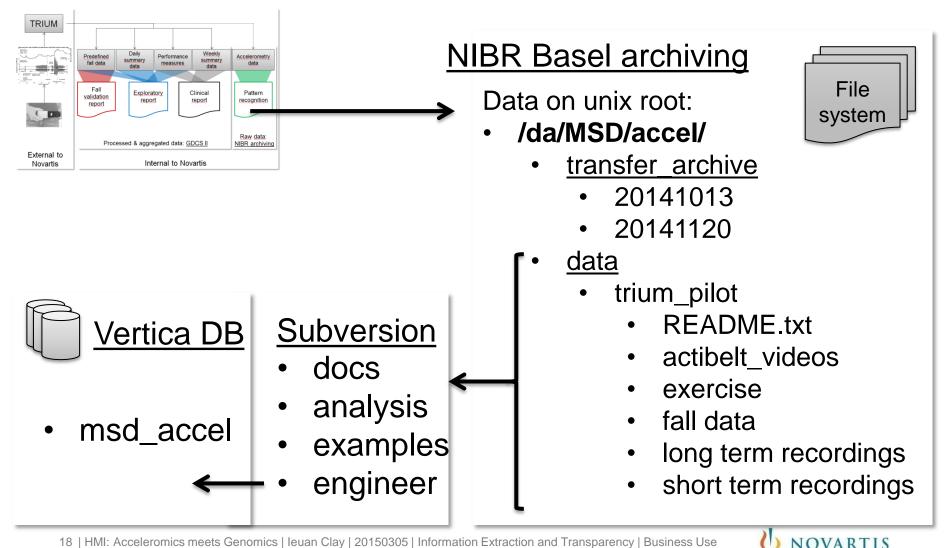




Peer| Preprints | https://doi.org/10.7287/peerj.preprints.2546v1 | CC BY 4.0 Open Access | rec: 21 Oct 2016, publ: 21 Oct 2016

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Data Flow III: Archiving of exploratory data

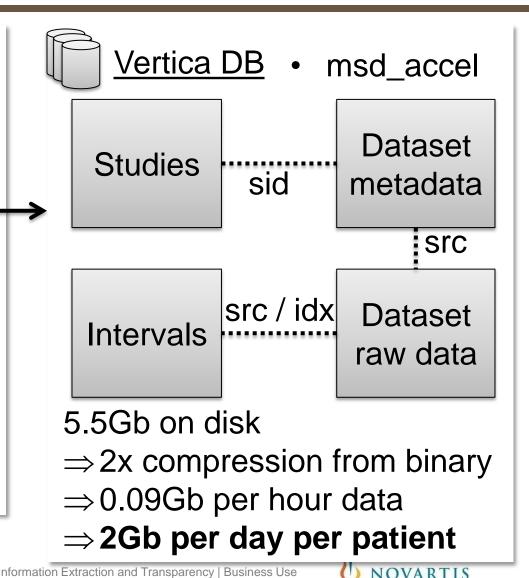


Data overview: Database structure Data storage

File system

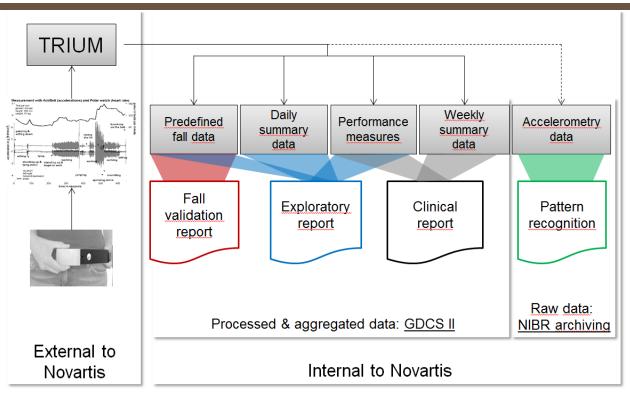
file	BIN	FLAT
54256DDB.BIN	117M	1.4G
MDfall20110901.bin	1.1M	15M
1408001464.bin	150M	1.5G
1408396914.bin	512M	6.1G
1408396625.bin	62M	748M
1408555060.bin	449M	5.3G
53318EC1.BIN	1.4G	22G
54368376.BIN	39M	594M
1408001464a.bin	150M	1.5G

64 hours of data ~2.9Gb binary data ~40Gb 'flat' data =>15x (de-)compression



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Data Flow IV: Division within Novartis



Aims ("Added Value")

Through deeper understanding of the physical activity of patients, we aim to deliver:

- Better patient selection
- Simpler trials and reduced patient burden
- Better outcome measures

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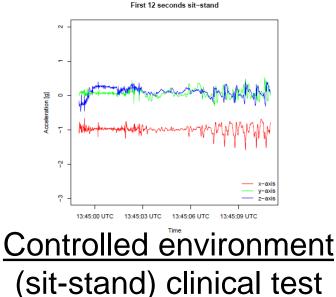
Analytic challenges

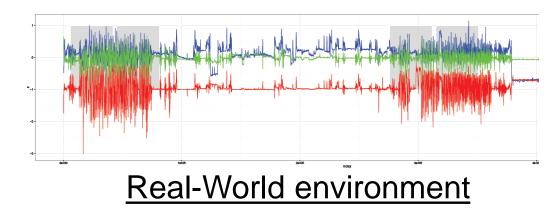
Aims ("Added Value")

Through deeper understanding of the physical activity of patients, we aim to deliver:

- Better patient selection
- Simpler trials and reduced patient burden
- Better outcome measures
- Challenges:
 - Data QC (What is normal?)
 - Real-world, streaming data (Variability & complexity)





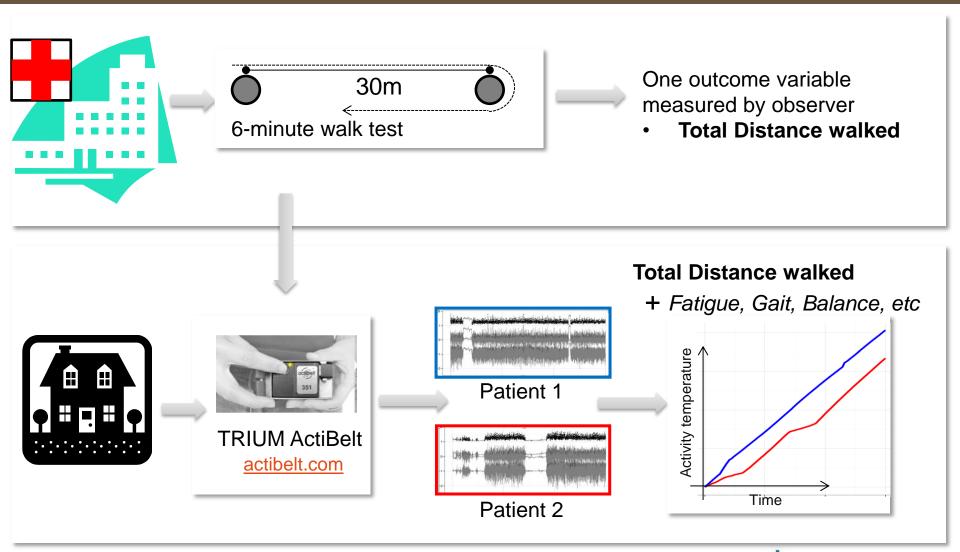


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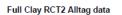
6-minute walk test

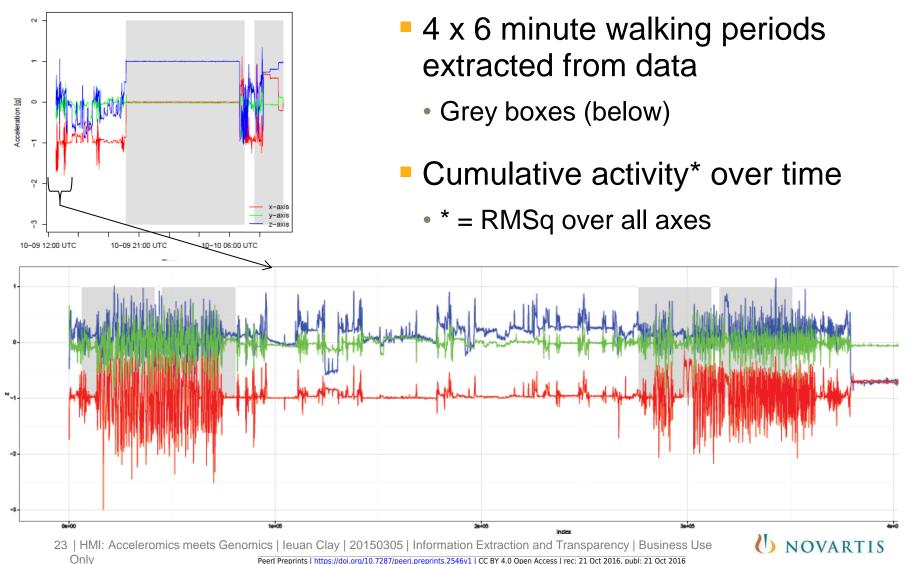
Current gold-standard



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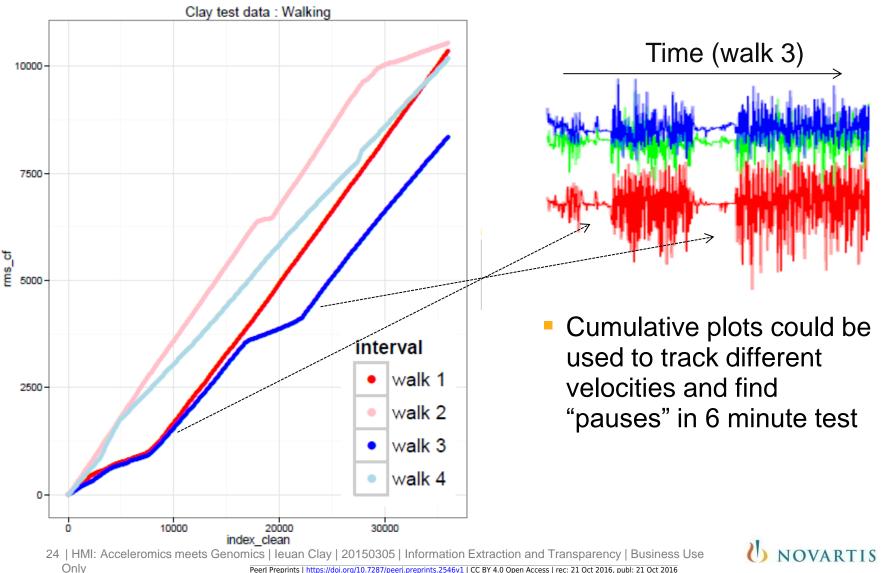
6-minute walk test II Exercise (Clay "6 minute walk")





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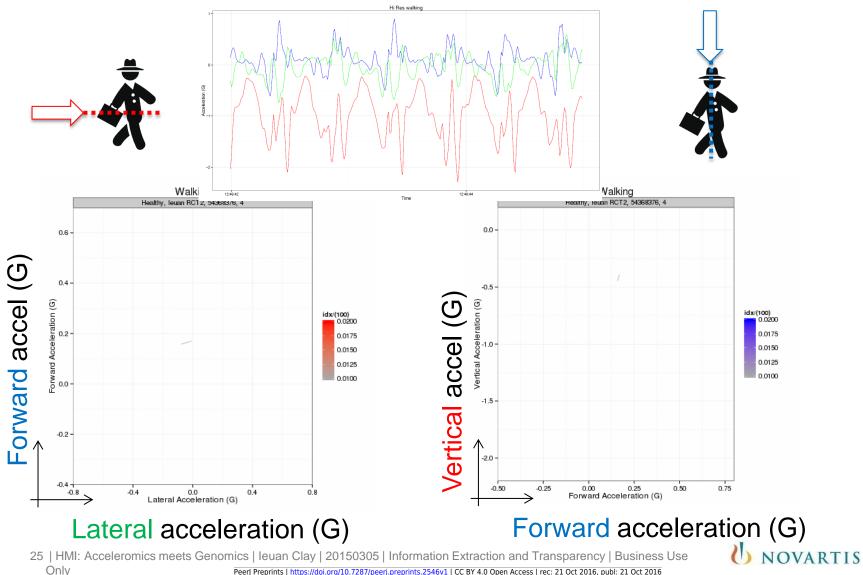
6-minute walk test III Exercise (Clay "6 minute walk")



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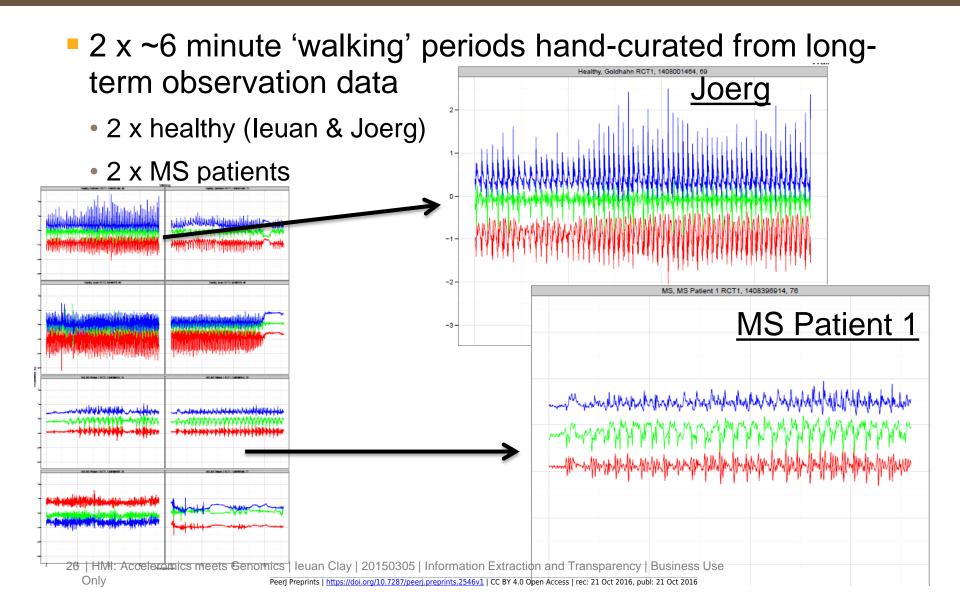
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Walking assymetry (gait 'quality')

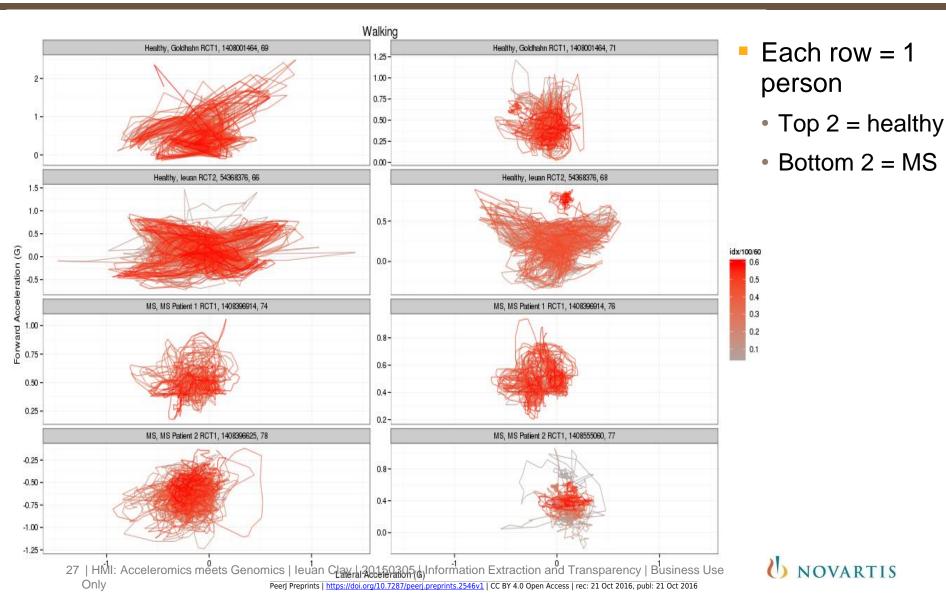


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Gait II Real life or 6MW



Gait II Variation within and between individuals

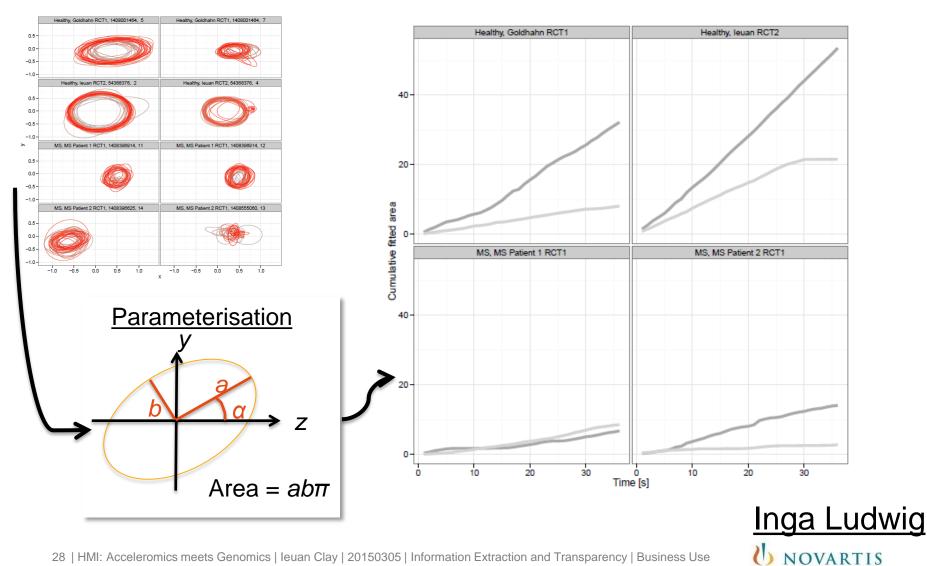


NOT PEEE Each panel = 1 person

• Top 2 = healthy

Gait II Variation within and between individuals

Bottom 2 = MS



Outlook

- The possibilities for information extraction are huge
- Documentation and other best practices are crucial to enabling analysis
- The tools exist
- Will bigger data = bigger problems (scaling infrastructure)?



Discussion

Questions, suggestions, feedback!



- Jörg Goldhahn, Daniel Rooks, Sophie Lemire-Brachat, Valerie Lanctin, Miriam Porter, Thomas Vogel, Henning Schmidt, Inga Ludwig, Lorcan Walsh, Thierry Clade, Nick Holway, Ieuan Clay
- Collaborators at TRIUM / The Human Motion Institue
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