



Challenges to GCP Compliant Deployment of Biosensors in Cognitively Impaired Populations

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Challenges to GCP Compliant Deployment of Biosensors in Cognitively Impaired Populations

AGENDA

- IXICO overview
- Overview of challenges
 - Ethical, privacy and security
 - Operational
 - Data validation
 - Authentication
 - Audit trail
 - Computer system validation
 - Data quality control
- Case studies of deployment
 - Cygnus study
 - Context study

IXICO Overview

- ▶ **IXICO brings innovative digital technologies and services to those involved in researching and treating diseases of the brain**
- ▶ **IXICO's primary focus is on the needs of the pharmaceutical industry to support**
 - Development of drugs in clinical trials
 - Use of drugs in the clinic to improve patient safety and outcomes
- ▶ **Our background is in imaging-derived biomarkers for enrichment, safety and efficacy**
 - The evolving biosensor-derived biomarkers landscape parallels imaging 15-20 years ago
- ▶ **Acquisition of Optimal Medicine in 2015 provides new capabilities and resources to apply to biosensors**
- ▶ **Collaborations with Withings/Nokia and Prof Yves Dauvilliers, Montpellier University Hospital Sleep Lab, DPUK**



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Ethics, Privacy and Security

- ▶ **Biosensors worn at home introduce particular challenges**
- ▶ **These increase if:**
 - The data are accessed or shared within a study
 - Real-time data are collection
 - Using consumer devices via the cloud
 - Patients are cognitively impaired
- ▶ **IXICO is particularly focussing on these issues in Cygnus**
 - Protocol, ethics approval and informed consent challenges addressed with UK authorities and patient groups

Overview of Challenges to GCP Compliant Deployment

- ▶ The challenges depend on the context of use
- ▶ Illustrated with two case studies from IXICO



- To evaluate feasibility of biosensor & mobile technologies for collecting high quality and actionable data from cognitively impaired patients and their care givers
- Real world setting
- Linking to standard clinical care dataset about care delivery, cost and outcomes
- 250 patients presenting to memory services and 250 care givers
- Sub-group with wearables





- To explore the performance of biosensors compared with standard sleep assessment methods (polysomnography, sleep scales and sleep diaries) in healthy elderly subjects and individuals with neurodegenerative disorders
- To discover clinically valid actigraphy-derived sleep metrics (and beyond) that take into account physiology and pathology
- Commencing with elderly control subjects and Parkinson's disease

Device Selection For Cygnus

▶ Device evaluation

- Detailed assessment of a number of devices on parameters key to adoption and study success

	Price	Comfort	Step Count	Clock	Raw data	Ease of use	Data access	Battery life	Aesthetics
 Fitbit Charge	80	Good	Yes	Yes	No	OK	Web	5 days	OK
Garmin Vivofit	60	Good	Yes	Yes	No	OK	Web	1 year	OK
Axivity	150	Poor	Yes	No	Yes	Good	Raw	30 days	Poor
 wGT3X-BT Monitor	300+	?	with process	No	Yes	?	Raw	25 days	Poor
Withings Activite Pop	100	Good	Yes	Yes	No?	OK	API	8 months	Good
ActiGraph GT9X	180	Poor	Yes	Yes	Yes	OK	Raw	7 days	Poor



▶ Focus groups

- Highest scoring devices were evaluated in focus groups of people diagnosed with mild cognitive impairment/dementia and their care-givers

Practical Considerations: Trade-offs

- ▶ Long battery life and highly processed data vs. short battery life and raw data
- ▶ No charging worn 24/7 for battery life vs. removed for charging with data gaps
- ▶ Familiar appearance (watch) of consumer device but processed data



- Withings Activité Pop
- 8 months battery life
- Processed data
- Requires app and website
- Syncs daily
- Near real-time QC of data
- Data can be “actioned” during study (ePRO questions, troubleshooting, technology issues...)

CONTEXT

Adding perspective to biosensor data



- Axivity AX3
- 14 days battery/data collection (depends on settings)
- Raw data
- Standalone device
- Data download, viewing and QC only at end of study period
- Subject behaviour not influenced by viewing data

→ Three devices being used at present; Withings Activité Pop, Axivity and Actigraph Link

▶ Authentication

- How do we know who the data are coming from?

▶ Audit trail

- Especially for consumer devices, how do we identify and document changes to data made by people or systems?

▶ Data quality control

- Real-time verification e.g. syncing
- Artefact detection
 - Not being worn
 - Noise and error
 - ...

→ Risk based approach to computer system validation

Audit Trail and Computer System Validation For Consumer Devices

▶ How do I know changes in sensor measurement are due to change in subject rather than change in sensor, its software, or artefact?

- Reliable data
- Regulatory compliance for decision making



▶ Traditional approach


- Vendor assessment of wearable supplier
- Validation of computer system (audit trail, access etc)
- Validate consumer wearable (inc. all systems) for “context of use” against “gold standard”
- Formal change control process if changes in device or software


Data QC - Near Real-Time Verification


Participants Welcome Janet.Munro


		First Name	Last Name	UPI	Date Of Birth	Type	Consent Date	
✓	!	*****	*****	02_PA_014	1944-06-01	Patient	2016-09-27	Edit
		*****	*****	02_SP_014	1954-12-28	Study Partner	2016-09-27	Edit
		*****	*****	03_PA_008	1932-03-04	Patient	2016-08-28	Edit
✓	✓	*****	*****	02_PA_016	1954-07-08	Patient	2016-09-29	Edit




***** 02_PA_016 

 Activity last 14 days Last sync at 17-Oct-2016 08:58:31

 ***** made 38 steps today and 3059 yesterday [Detail](#)

 Sleep last 14 days Last sync at 17-Oct-2016 04:45:08

 [Detail](#)



 Sleep last 14 days

Show Sleep Measures

Date Description	Day	Date	Duration To Sleep	Light Sleep Duration	Deep Sleep Duration	Wakeup Duration	Wakeup Count
today	Mon	17-Oct-2016	00h 01m	02h 39m	01h 02m	00h 02m	1
yesterday	Sun	16-Oct-2016	00h 00m	01h 44m	02h 11m	00h 26m	3
2 days ago	Sat	15-Oct-2016	00h 00m	04h 16m	02h 10m	00h 18m	3
3 days ago	Fri	14-Oct-2016	00h 00m	02h 11m	03h 05m	00h 16m	2
4 days ago	Thu	13-Oct-2016	00h 04m	02h 59m	05h 37m	00h 40m	3

Challenges to GCP Compliant Deployment of Biosensors in Cognitively Impaired Populations

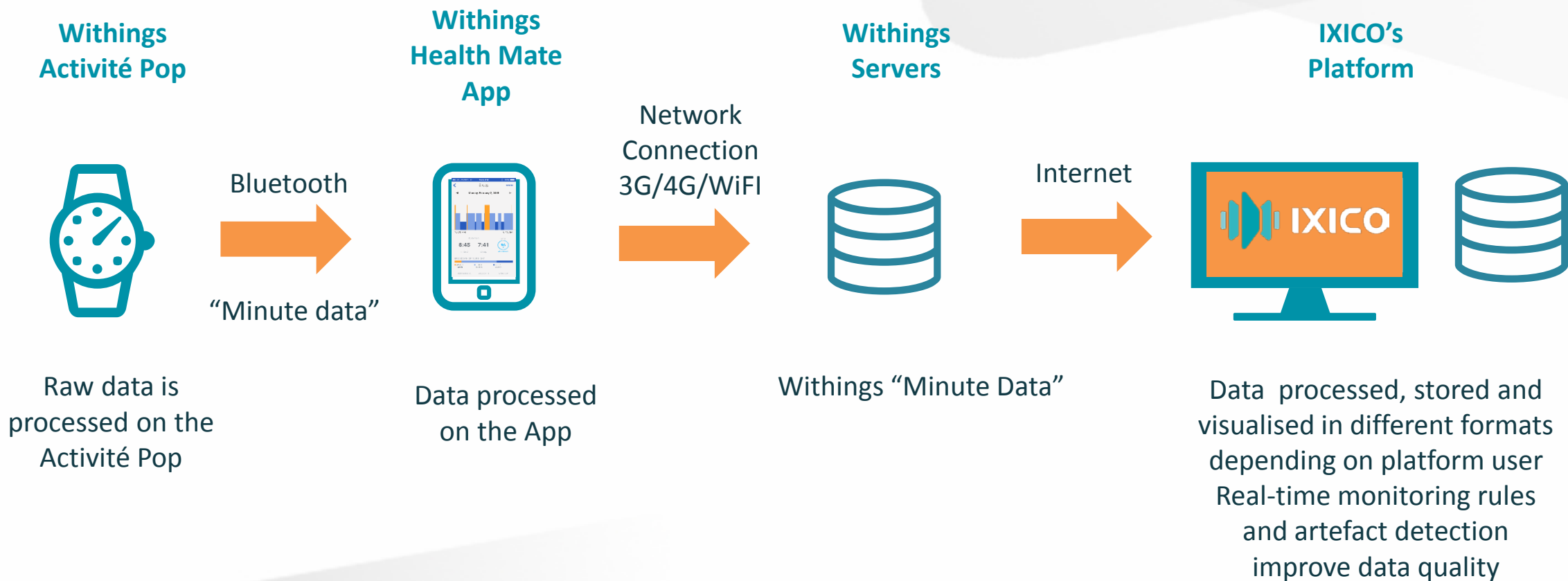
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Operationalising Wearables in the GCP Environment

- ▶ **Focus groups completed**
- ▶ **Ethics approval obtained**
- ▶ **Standard Operating Procedures established**
- ▶ **GCP compliant software platform with integrated wearables, ePRO and ObsPro (remote monitoring from patients and caregivers partners)**
- ▶ **>170 subjects recruited in main study**
- ▶ **Wearable sub-study data collection commenced**
- ▶ **Data artefact and syncing processes operational**

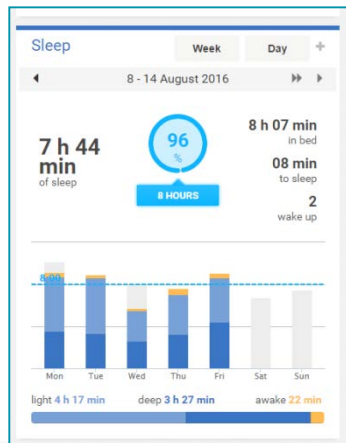
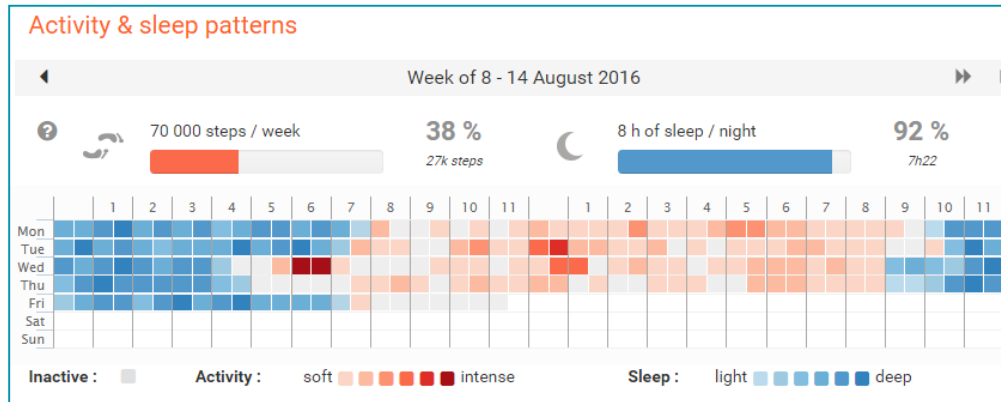
Dataflow from Withings Activité Pop to IXICO's Platform



Sleep Measures from "Minute Data"

▶ Withings "consumer" data

"Heat-maps" on Withings' website



Graphs on Withings' App

▶ Withings data from Oauth API

- Example of detailed sleep data

```

{
  "id":16616514,
  "timezone":"\\"Europe/Paris\\" ",
  "model":32,
  "startdate":1410521659,
  "enddate":1410542577,
  "date":"2014-09-11",
  "data":
  {
    "wakeuptime":1800,
    "lightsleepduration":18540,
    "deepsleepduration":8460,
    "remsleepduration":10460,
    "durationtosleep":420,
    "durationtowakeup":360,
    "wakeuptime":3
  }
}
    
```



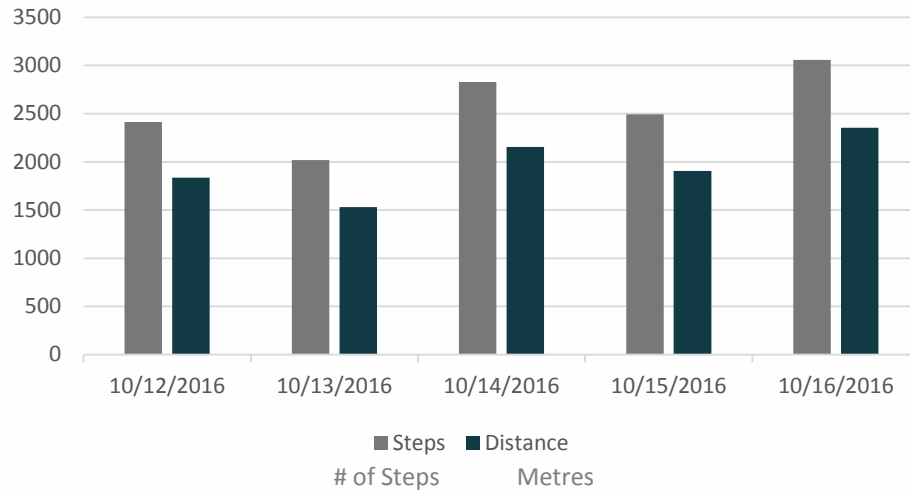
Sample output converted to excel

state	Start time	End time	Start date	End date
awake	20:41:00	20:46:00	31/07/2016	31/07/2016
light sleep	20:46:00	20:49:00	31/07/2016	31/07/2016
deep sleep	20:49:00	21:09:00	31/07/2016	31/07/2016
light sleep	21:09:00	21:18:00	31/07/2016	31/07/2016
awake	21:18:00	21:30:00	31/07/2016	31/07/2016
light sleep	21:30:00	21:51:00	31/07/2016	31/07/2016
deep sleep	21:51:00	23:34:00	31/07/2016	31/07/2016
light sleep	23:34:00	23:45:00	31/07/2016	31/07/2016
awake	23:45:00	23:49:00	31/07/2016	31/07/2016
light sleep	23:49:00	23:58:00	31/07/2016	31/07/2016

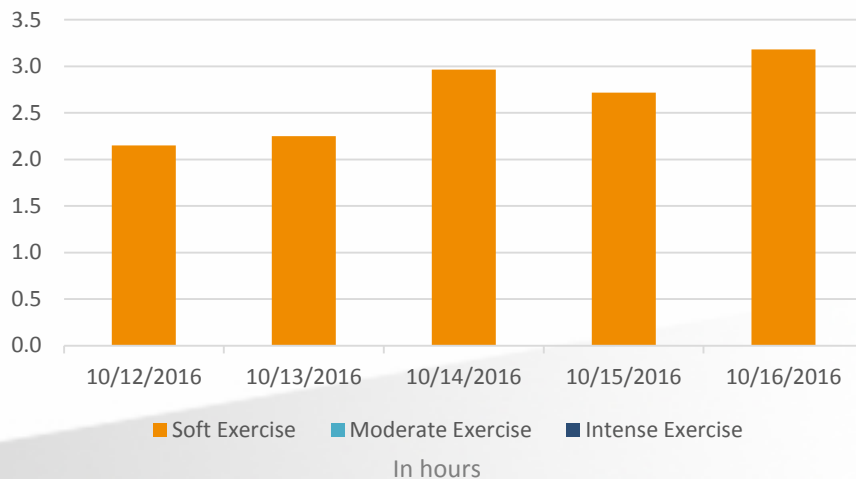
API "Get sleep data summary"

Subject CY-01 “Minute data” Summaries

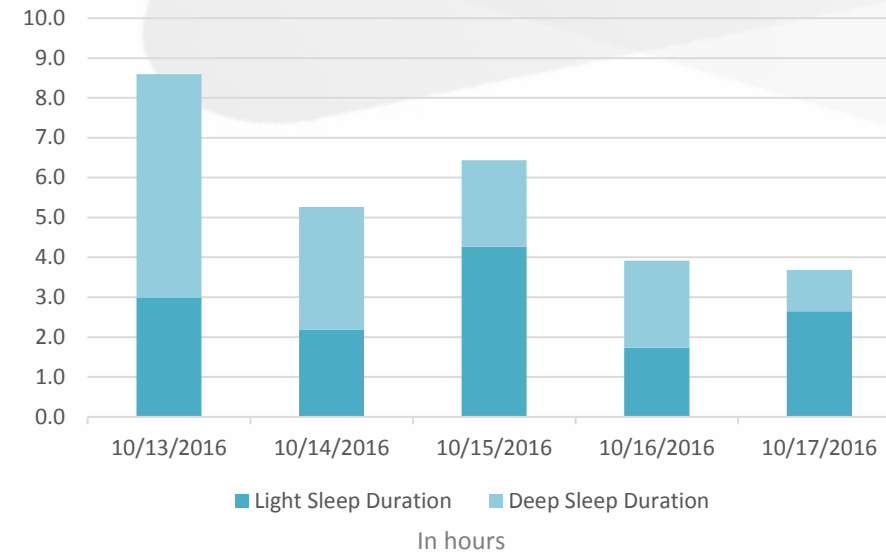
Activity Summary



Intensity of exercise



Sleep Summary

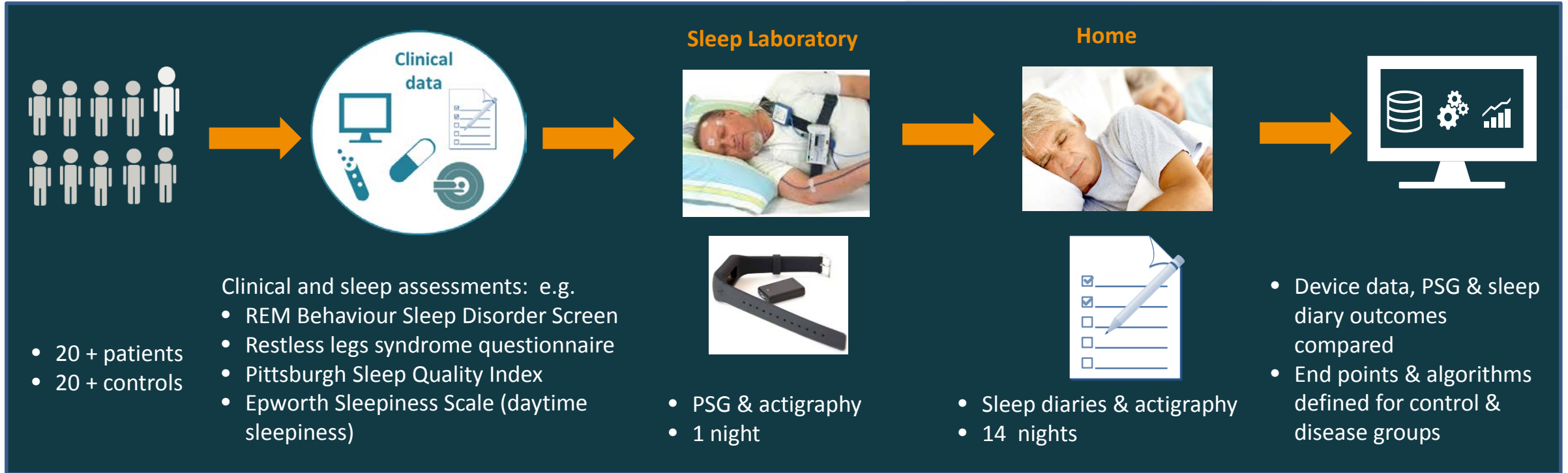


- ▶ 63 year old female
- ▶ Early onset Alzheimer’s Disease

Case Study: Enabling Context-Appropriate Use and Interpretation of Biosensor Data

- ▶ **Advancing the understanding of the impact of clinical context on biosensor data**
 - To explore the performance of biosensors compared with standard sleep assessment methods (polysomnography, sleep scales and sleep diaries) in healthy elderly subjects and individuals with neurodegenerative disorders
 - To discover clinically valid actigraphy-derived sleep metrics that take into account physiology and pathology
 - To identify bio-signatures of sleep disorders that influence sleep interpretation (REM- or sleep breathing-disorders, restless leg, etc)
- ▶ **Understanding complementary roles of different sleep assessment methods**
- ▶ **>15 subjects recruited; elderly control subjects (with detailed cognitive and clinical assessments) and Parkinson's disease patients**

Actigraphy vs Polysomnography PSG and Sleep Diaries



CONTEXT

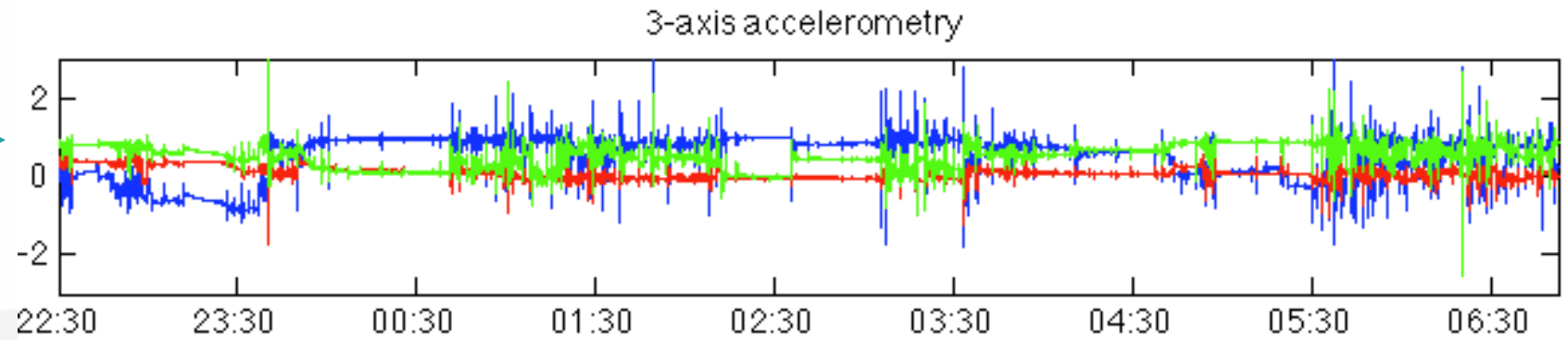
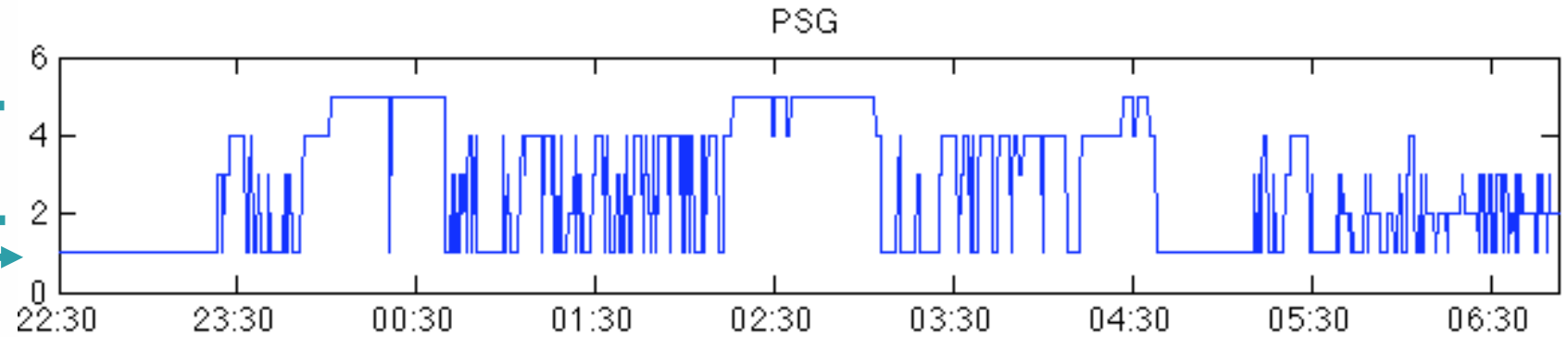
- ▶ 71-year old female
- ▶ Parkinson's disease
- ▶ 1 night PSG plus Actigraphy

Subject PD-1 "Raw data"

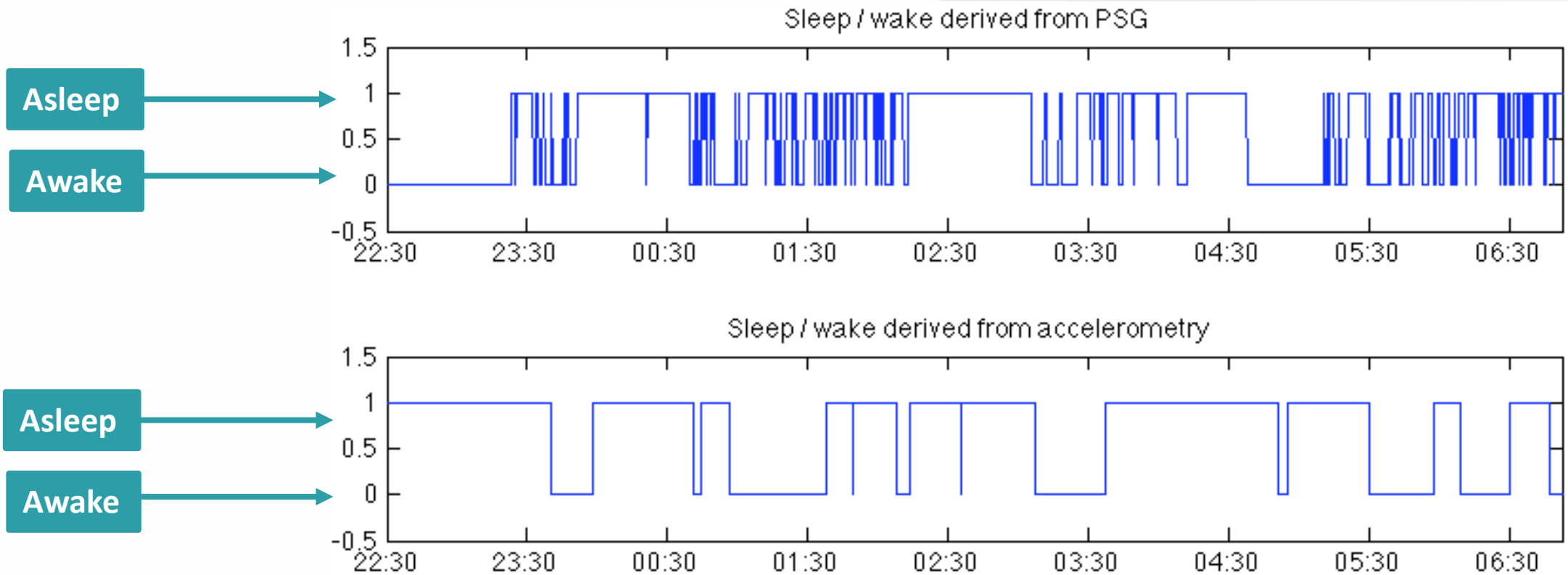
Four sleep stages

Awake

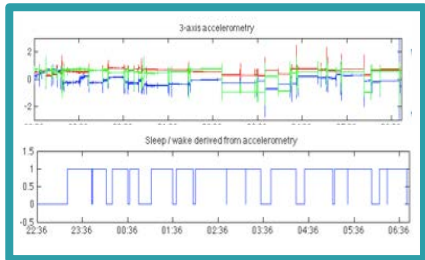
Raw accelerometer
Blue: x-axis
Red: y-axis
Green: z-axis



→ Good correlation between PSG and raw actigraphy



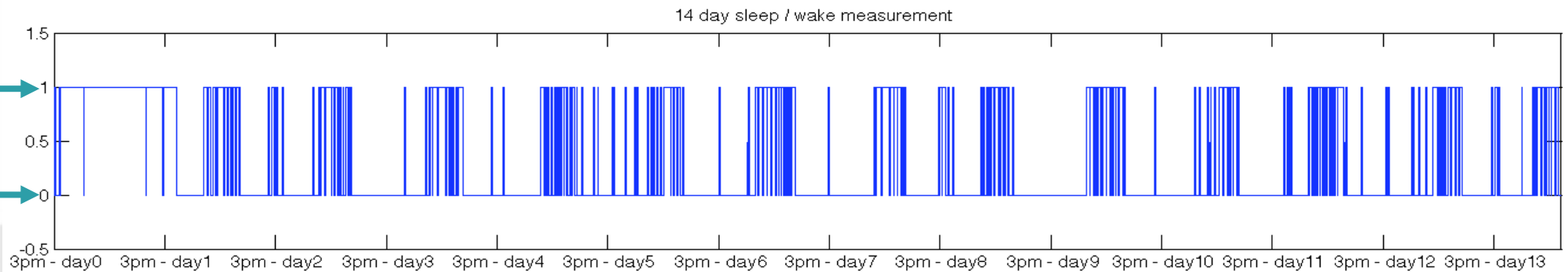
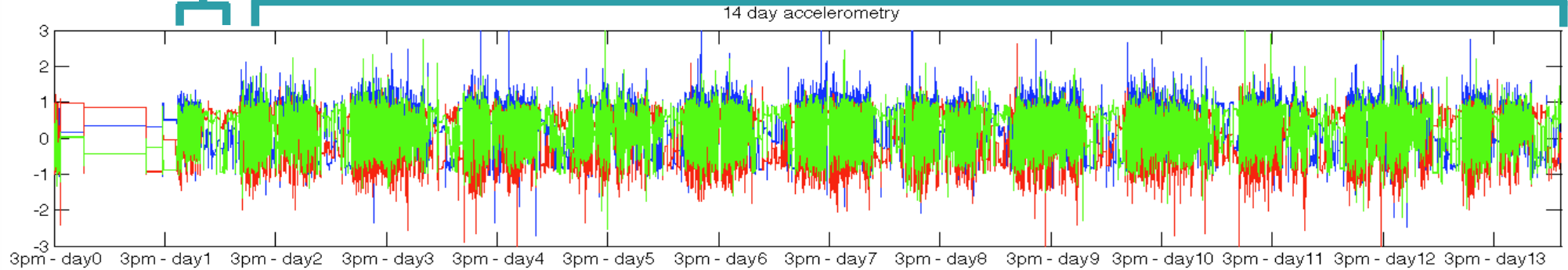
→ Poor correlation between PSG and “processed” actigraphy using standard sleep algorithms



**1 night sleep lab recording
PSG & actigraphy**

**14 night home recording
Sleep diary & actigraphy**

1. Hier, j'ai fait le sommeil entre _____ et _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
2. Hier, j'ai pu _____ rep. de m'endormir vite _____ en d'abord pour me coucher à _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
3. Je me suis réveillé à _____ mais si j'ai dormi les dernières à _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
4. Après avoir fermé les yeux, je me suis endormi(e) en _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
5. Hier, j'ai pu _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
6. Hier, j'ai pu _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
7. Ce matin, je me suis réveillé(e) à _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
8. Au cours de la nuit, je me suis réveillé(e) _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
9. Au cours de la nuit, je me suis réveillé(e) _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00
10. Hier, j'ai pu _____	14:20	14:25	14:30	14:35	14:40	14:45	14:50	14:55	15:00	15:05	15:10	15:15	15:20	15:25	15:30	15:35	15:40	15:45	15:50	15:55	16:00



1 = asleep

0 = awake

Thank You