

Data-driven Digital Healthcare: Longitudinal Data Analytics for Parkinson's Disease

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- cloudUPDRS certified Class I Medical Device for clinical use
- Rate PD motor symptoms as precisely as an experienced clinician
- Extends and adapts Part III of the standard UPDRS protocol
- Unsupervised use at home
 - Employs accelerometer for tremor and gait measurements
 - Employs touch-screen for tapping measurements
 - No clinical or technical supervision during testing: bespoke user journey
- Data analytics



- 1. Ensure unsupervised test is carried out correctly
- 2. Reduce testing time
- 3. Capture symptom variability
- 4. Identify high quality signal segments

People with Parkinson's







Parkinson's Disease (PD)

- No cure
- Managed mainly by replacing dopamine
- Motor symptoms
 - tremor, rigidity, slowness of movement (bradykinesia), freezing of gait, stiffness, shaking, falls
- Non-motor symptoms
 - bladder, memory, sleep, addictive behaviour, fatigue, pain, hallucinations





Symptoms of Parkinson's Disease

resting tremor

parkinson's disease

drcrunch.co.uk





Disease Progression





UNIFIED PARKINSON'S DISEASE RATING SCALE

- Standard clinical protocol for assessing PD
- Part III clinical assessment of motor symptoms
- Known issues:
 - time intensive
 - inter-rater variability
 - not sensitive
- Used formally
 - Drug trials
 - Consideration of advanced therapies
- Used informally as part of clinical assessment (once or twice per year) of disease progression
- Can we replace with an app?

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0: Normal:	No rigidity.	
1: Slight:	Rigidity only detected with activation maneuver.	
2: Mid:	Rigidity detected without the activation maneuver, but full range of motion is easily achieved.	RUE
3: Moderate:	Rigidity detected without the activation maneuver; full range of motion is achieved with effort.	\square
4: Severe:	Rigidity detected without the activation maneuver and full range of motion not achieved.	we
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Expressed use intentions

- Quantitative and qualitative methods
 - survey and audience panels
- Majority of PD patients would use app (86%)
 - Most would prefer the test to last less than 5 minutes per assessment (64%)
 - Some would accept up to 10 minutes (27%)
- Main motivation: Need to understand their condition
- No expressed privacy concerns

cloudUPDRS app



- Design objectives:
 - Sensitive to patient mobility constraints
 - Sensitive to patient cognitive impairments
- Approach:
 - Constrain user context for reliable interpretation of data
 - Encourage frequent use



CONTINUE

Test Movements





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Architecture



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Clinical dashboard



Data Processing Pipeline







Example: Tremor processing pipeline

From raw accelerometer data to UPDRS score (0-4 scale)



Open Source PDkit for python on github



CUSSP Clinical Study





- Achieve firm user adherence to the prescribed movements
 - Accept test record only when movement executed correctly
 - Reject test when movement does not match expectations
- Use deep learning to learn movement features
- Apply offline or online (i.e. at the server on in the app)
- Use Tensorflow to learn and apply model



Tremor signal



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Deep Learning Architecture









Performance using DNN

Classifiers	Accuracy	F1-score	AUC	
ExtraTrees	0.73	0.79	0.83	
BernoulliNB	0.73	0.79	0.83	
RandomForest	0.73	0.79	0.83	
$\operatorname{GradientBoosting}$	0.72	0.80	0.83	
Bagging	0.72	0.78	0.83	
AdaBoost	0.66	0.75	0.81	
GaussianNB	0.69	0.75	0.83	
DMLP	0.75	0.81	0.85	
RCNN	0.78	0.82	0.87	

	TP	(%)	FN	(%)	TN	(%)	FP	(%)
ExtraTrees	141.52	93	8.98	6	13.36	17	63.14	82
BernoulliNB	146.23	96	4.27	3	6.92	8	69.58	91
RandomForest	138.39	91	12.11	8	16.19	20	60.31	79
GradientBoosting	146.02	96	4.48	3	8.12	10	68.38	89
Bagging	135.58	89	14.92	10	18.03	23	58.47	76
AdaBoost	128.0	84	22.5	15	17.34	22	59.16	77
GaussianNB	116.01	76	34.49	23	35.41	45	41.09	54
DMLP	135.73	89	15.77	10	28.19	37	49.31	63
RCNN	133.22	87	18.28	12	38.38	50	39.12	50



- UPDRS exhaustive search of all possible symptoms
- Each patient presents only a few
- Symptoms typically change slowly e.g. 6 months
- ~6 features are predictive of overall score
- Use ML to identify the specific tests that offer the highest inferential power
 - Observer five full tests
 - Apply standard ensemble of randomized decision tree method to rank tests according to predictive strength
 - Select top 3 tests for individualised quick test





- Minimum Detectable Change (MDC95) ~12 (range 0-108)
- Typical annual disease progression 3-4 points
- Idealised response model
- Rapid uphill, slow downhill
- Affected by numerous parameters e.g. mood, social interaction, diet, exercise etc
- One sample has extremely limited value



Capturing temporal variability

- Tremor signal is not stationary but is often treated as such (not unreasonable due to measurement limitations)
- Consider tremor to be a random process
- Look at temporal aggregates
- Preliminary results suggest far superior MDC95













Gait test/activity recognition with HMM



How precisely can we detect the onset of turning movements? Healthy subjects turn differently than PwP.



- Move from clinician to automated diagnosis and treatment offers great opportunities to realise patient benefits
- Challenges often relate to having to change methods
- This can be intensified by the greater availability of data
- Stationary to dynamic processes, non-linearity
- Validated evidence is time consuming/expensive to collect



Further resources

App demo videos

http://www.updrs.net

PDkit analytics toolkit

https://github.com/pdkit/pdkit

Papers

http://www.dcs.bbk.ac.uk/~gr/pubs.html

CUSSP Study Record

https://clinicaltrials.gov/ct2/show/NCT02937324

cloudUPDRS app on the Play Store

https://play.google.com/store/apps/details?id=uk.ac.bbk.dcs.cloudupdrs



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