2017 춘계심혈관통합학술대회

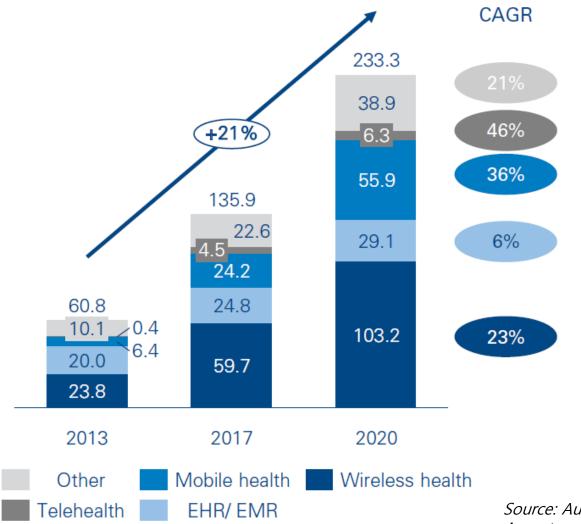
Application of Dr.M, Integrated Mobile Healthcare Strategy, for Treatment of Cardiovascular Disease

doceitore



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Digital Healthcare Market (\$B)

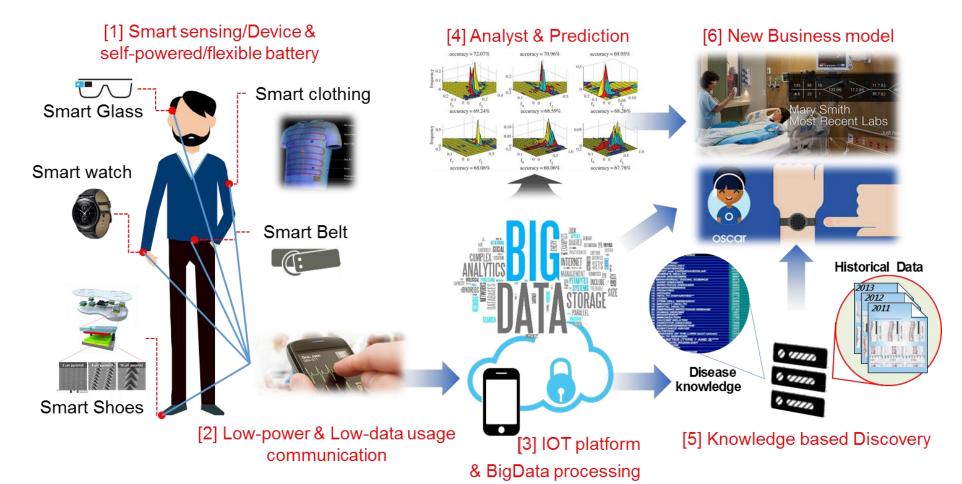


Source: Author D. Little, GSMA, Allied Market Research, Accenture, IHS, MarketsandMarkets

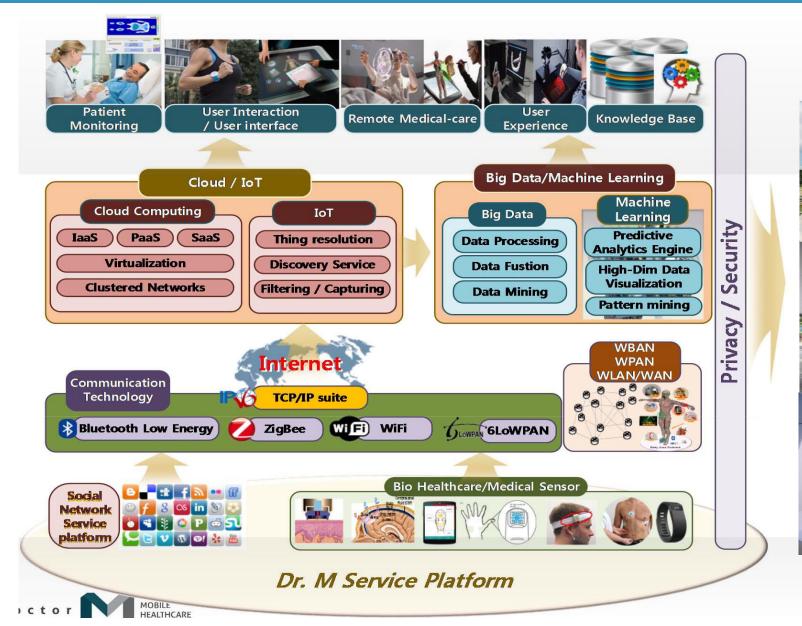
Dr.M-mobile healthcare innovation @KAIST

Dr.M Platform aims for the integrated mobile healthcare service

- ITC technology + Medical service ⇒ **Personalized Healthcare (or Wellness) service**
- Real-time measure/transmission/cloud based storage/analysis



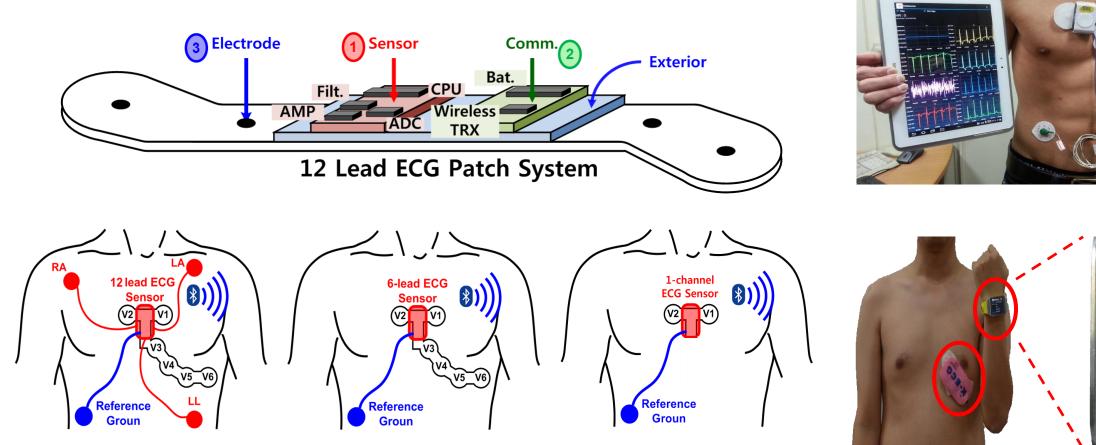
Classify of the Dr.M technology





K-PATCH: Mobile ECG PATCH

- Using wearable type of multi-channel ECG Patch
- ECG can be easily measured at anytime and anywhere



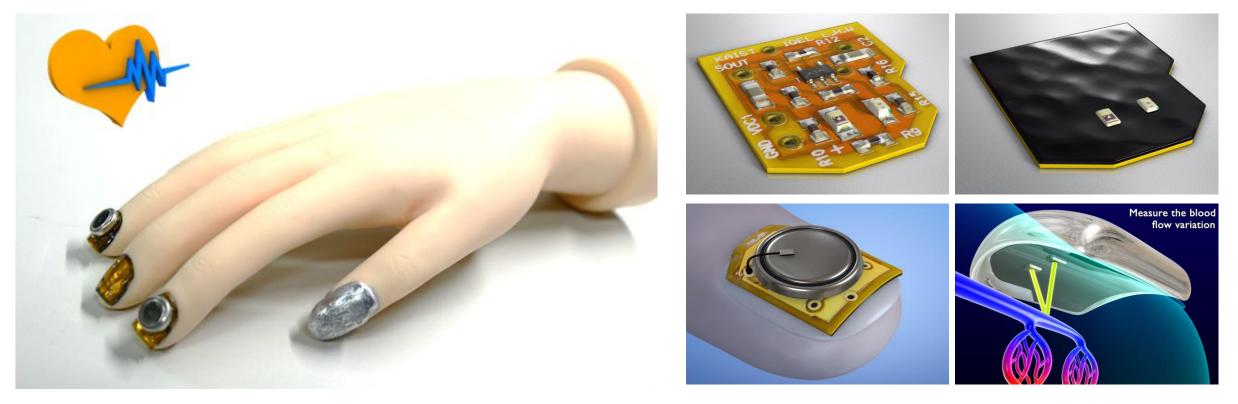
ECGMonitor

ple : 22068(177)Error : 0 Test : 145

HR:81

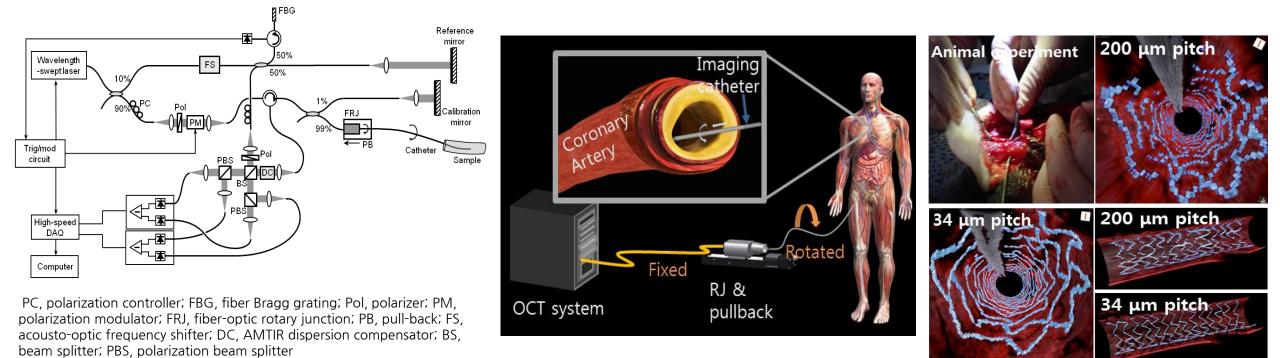
Thumbnail Heart-rate Monitors

• As the sensor is attached to the nail which is insensible, it is convenient for use ,and its conformal contact makes the measurement of the heart rate much easier.



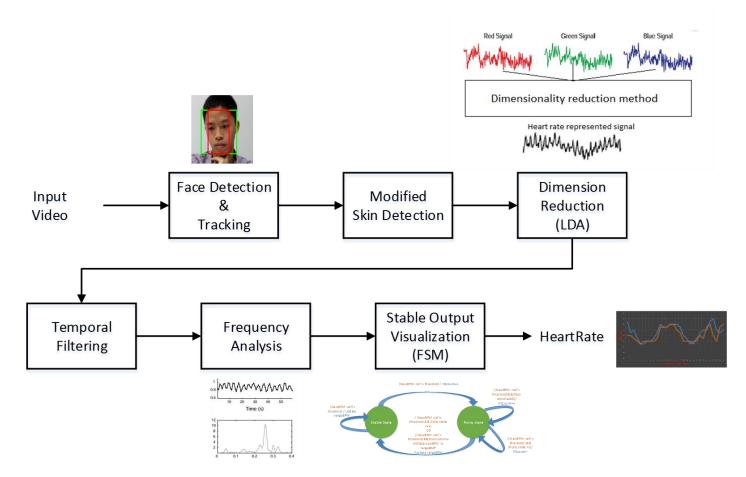
Intravascular OCT: High-speed scanning unit for cardiovascular OCT

- Development of high-speed rotary junction(RJ), imaging catheter and pullback stage for 3-D intravascular images
- To diagnose coronary artery disease accurately, high-speed and high-resolution intravascular OCT system is essential

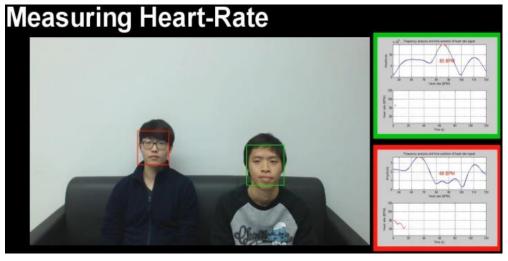


A Robust Real-time system for Heart-rate monitoring system via Camera

 Anyone who wants to monitor their heart-rate status non-invasively. Hospital can also use this software to monitor heart rate of patients remotely, especially for elderly people



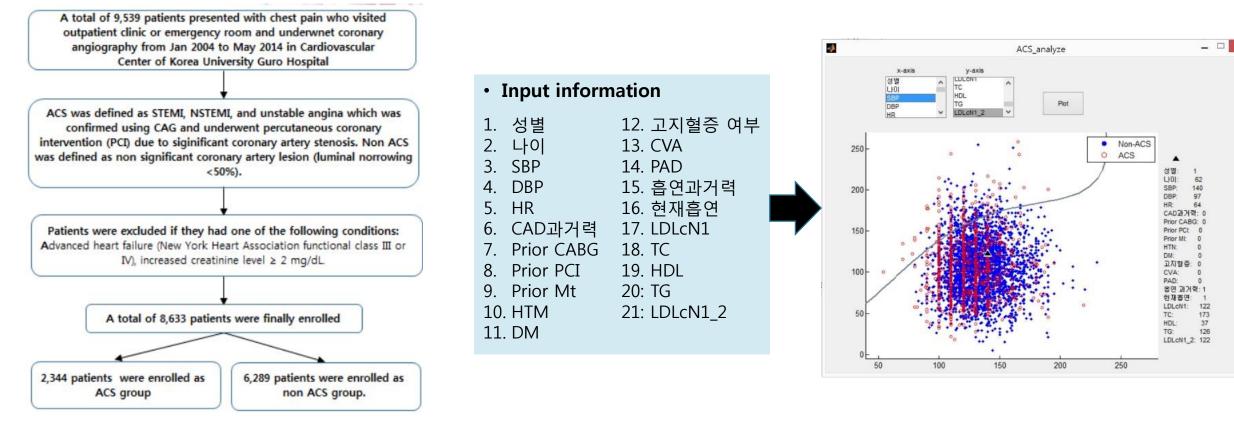




Acute Coronary Syndrome prediction & diagnosis

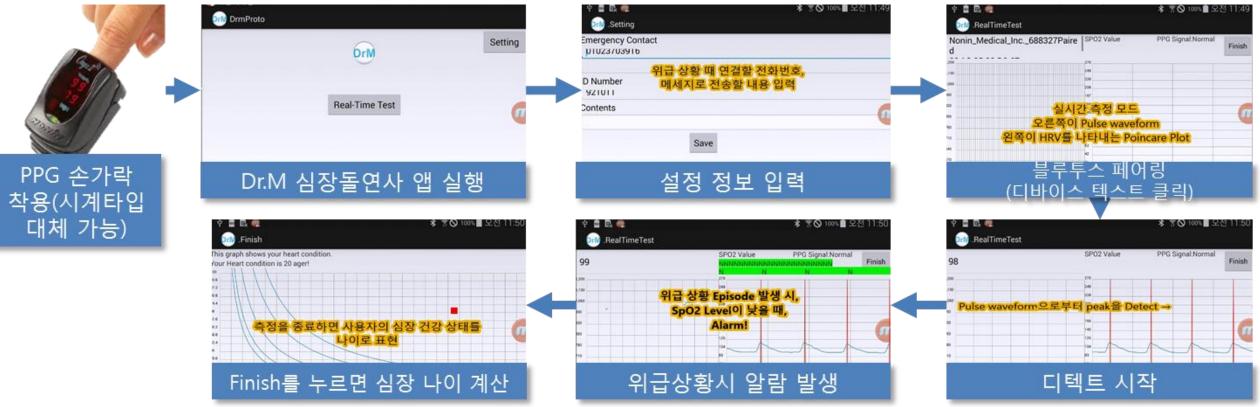
• Through Big Data analysis using machine learning algorithms, we support cardiologists' decision making to avoid

unnecessary coronary angiography.



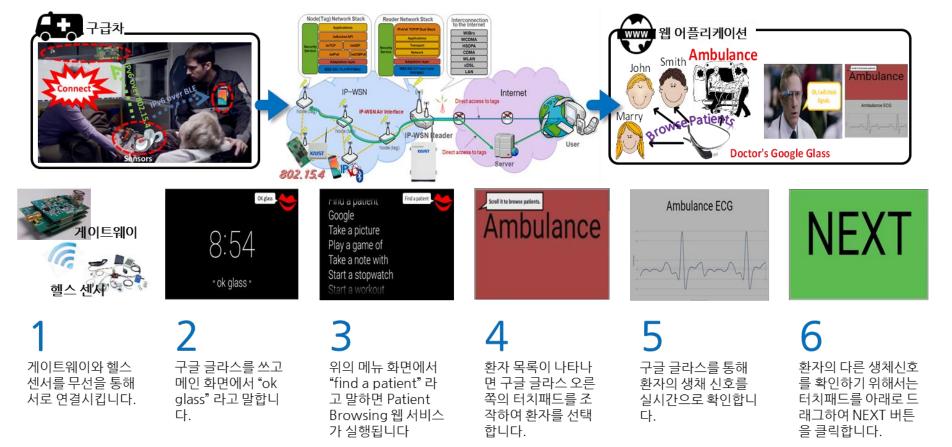
Secure Sensing and Network for sudden cardiac death (SCD)

• To develop the system preventing acute heart failure (AHF) of single residences who cannot adequately be helped in Gold Time.



SNAIL IPv6 IoT Connectivity Platform

- IoT connectivity platform, SNAIL : Sensor Network for an All Ip-world
- Patient browsing Scenario for real-time diagnosis and fast action
- Browse patient's health condition quickly using smart devices (i.e. Google glass.)







Dr.M research contents

Wearable Device & User

- Sensor & wearable device accuracy
- Sensor accuracy based on behavioral action (rest, walk, run.. Ect)
- User's personal behavior classification (typing, walking with smartphone, rest with smartphone)
- Smart Device Usage Pattern

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Individuals & Group & Community

- Behavioral changes in individual populations
- Big data analysis

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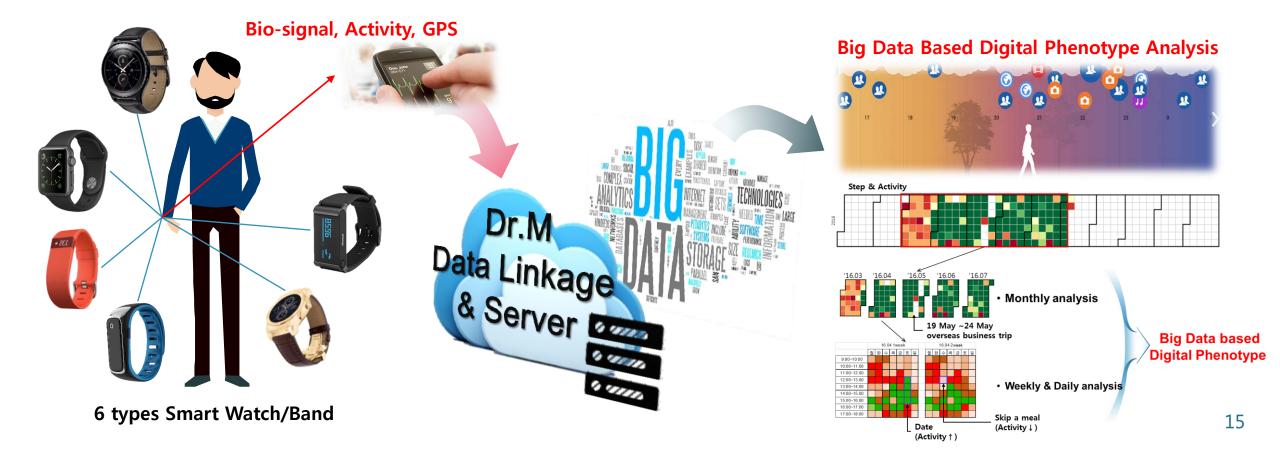
• Survey on device users

Personal Healthcare & Wellness service model

Utility of the Dr.M testbed

Utility of the Dr. M testbed for Digital Phenotype

- 6 types sensor device was distributed to 350 members of KAIST (April 2016 \sim)
- Real-time Sensor data (Heart rate, Activity, GPS) collection/monitor/analysis



Digital Phenotype

'Digital Phenotypes' describes the use of new digital technologies to capture individual data, outside of current healthcare settings.

 anything from wearable fitness devices and mobile phone apps, to social media and citizen science



Clinton Richard Dawkins (1941-03-26~) **The Extended Phenotype**: the Gene as the Unit of Selection. (Oxford: Oxford University Press; 1982) "phenotypes should not be limited just to biological processes, such as protein biosynthesis or tissue growth, but extended to include all effects that a gene has on its environment inside or outside of the body of the individual organism."

The digital phenotype

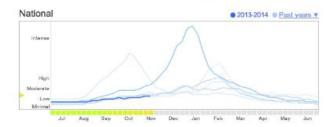
NATURE BIOTECHNOLOGY VOLUME 33 NUMBER 5 MAY 2015

Sachin H Jain, Brian W Powers, Jared B Hawkins & John S Brownstein

In the coming years, patient phenotypes captured to enhance health and wellness will extend to human interactions with digital technology.

Explore flu trends - United States

We've found that certain search terms are good indicators of flu activity. Google Flu Trends uses aggregated Google search data to estimate flu activity. Learn more a







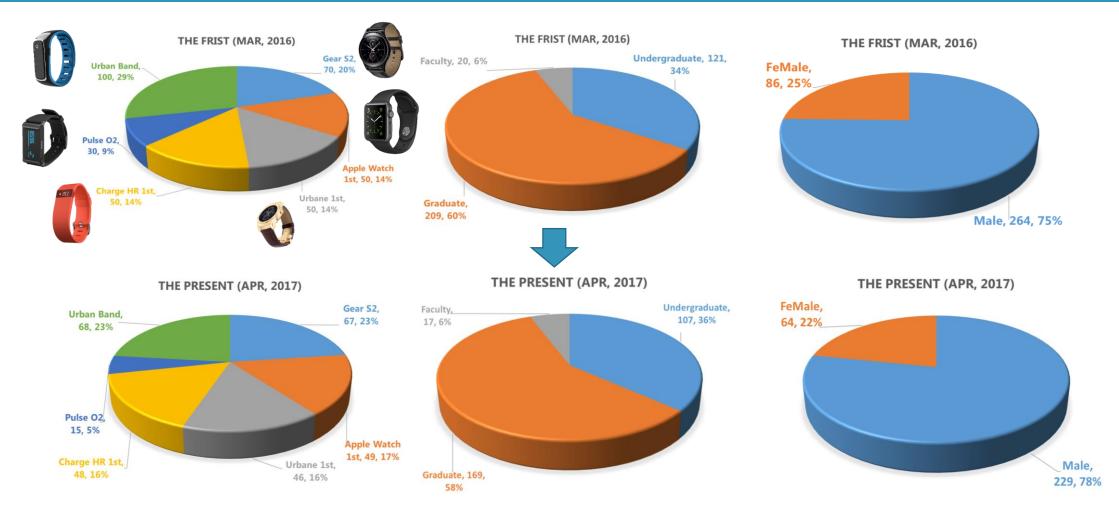
Digital Phenotype – Ginger.io

Your phone knows if you're depressed

• Time spent on smartphone and GPS location sensor data detect depression



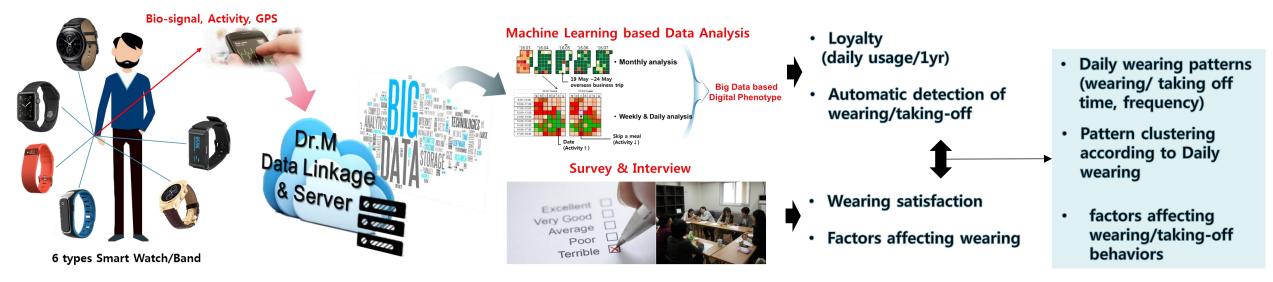
Utility of the Dr.M testbed



- 57 participants gave up: graduation, change jobs, device breakdown, inconvenience of use
 - 25 urban band user: breakdown of the device
 - 15 Pulse O2 user: inconvenience of use (Pulse O2 dose not support automatic Heart rate measurement)

Utility of the Dr.M testbed

Data Analysis, Survey, and Interview of Large-scaled Dr.M research



- Loyalty: Apple Watch>GearS2 = Charge HR > Urbane > Pulse O2 = Urban band
- Correlation with Satisfaction and Loyalty

Satisfaction	Correlation with Loyalty
Design	ho = 0.78
Notifications (call, text, email, alarm)	ho = 0.81
Health information	ho = 0.27
Useable time (battery)	ho = 0.43
Durability	ho = 0.72

Case study: wearing behaviors of Apple Watch

• Daily wearing patterns:

	•	
Average	MAARINA	hourc
Average	weamu	
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All days	10.48 hours (SD = 3.47)
Weekdays	11.32 hours (SD = 3.53)
Weekends	▶ 8.66 hours (SD = 3.60)

Don't want to be interrupted, and the costumes on weekdays and weekends are different, Longer wearing hours on weekdays than weekends

• Diurnal wearing pattern: 3 groups found from spectral clustering

				_	-					_		-								-				
		More wearing times										Less wearing times												
All-day wearer (35 people)		Working hours (9AM-midnight) (Mean=77.68%, SD = 4.04%)											 Off-hours (midnight-9AM) (Mean=65.85%, SD = 4.66%) 											
Working-hour wearer (7)		Working hours (9AM-midnight) (Mean=64.22%, SD = 5.82%)								Night-time (2AM-9AM) (Mean=36.60%, SD = 6.82%)														
Sleeping-hour wearer (4)		Night-time (2AM-9AM) (Mean=80.55%, SD = 10.56%)								 Working hours (9AM-midnight) (Mean=56.87%, SD = 3.36%) 														
All-day wearer Working-hour wearer		, ,	,		,	1	'	'		,	,	4		1	*			\ 	. * 	2	*			-
Sleeping-hour wearer	-			1			i.	i.		a.		1	4		1	Y.		à	1		4	1		-
	0	1	2	3	4	5	6	7	8	9	10	11 Tir	12 ne	13	14	15	16	17	18	19	20	21	22	23
Tak	eoff																							Wear

Users can separate the user into three groups depending on their wear time, which can be applied to the development of the service model

Case study: wearing behaviors of Apple Watch

8 factors affecting wearing/taking-off behaviors

- : Positive factor of wearing

- : Negative factor of taking-off

Factors	Affecting wearing behaviors	Affecting taking-off behaviors
Need of Immediate Response	 Ability to detect incoming notifications without missing Ability to quickly distinguish importance of notifications 	 No need of immediate response (In situations like off-working hours or weekends Desire to be away from notification alarms
Engagement of Social Activities	 Less disturbing others when checking information Glancing at watch is more socially acceptable action Less disturbing and easier to ignore alarm 	No relevant theme was found
 Availability of Personal Workspace 	Less likely to take-off smartwatch in shared spaces (e.g. café, restaurant)	More likely to take-off smartwatch in places with personal working area (e.g. office, classroom)
 Need for Multitasking 	 Activities with both hands occupied (e.g. driving a car, riding a bicycle) Activities with significant attention required (e.g. conducting an experiment) 	No relevant theme was found
 Charging Smartwatches 	No relevant theme was found	 Take-off smartwatch multiple times a day to charge when user is near charging stations Forget to wear smartwatch after charging Often forget or decide not to take charging stations (e.g. when traveling)
Aesthetics	More generous about the aesthetics as smartwatch is a 'wrist-worn electronic device'	 Smartwatch as a substitute of conventional watch (expectation as a smart fashion accessory)
Daily Activity Tracking	 Motivations from getting daily movement information Refresh from occasional physical movement status alarm 	 Disturbance from activity status update alarms Confusion with other useful notifications (e.g. emails or messengers)
• Exercise Tracking	 Desire to track and record their workout progress To keep track of their body status during workouts 	 Limited options of exercise types Unavailability to capture movements

Lessons so far...

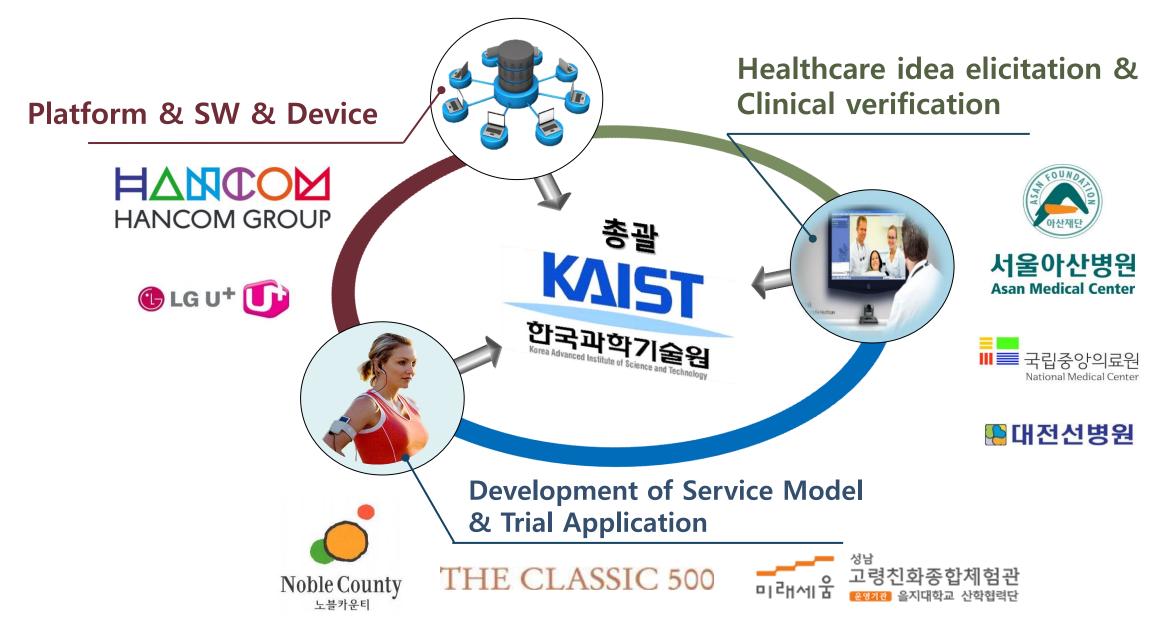
- Sensor accuracy, measured bio-signal (or information)
 - Inaccuracy of measured values (bio-signal, activity type, step count..)
 - Not enough current information (Heart-rate, Activity, GPS) to provide personal physical healthcare
- Lack of Healthcare function or service
 - Most user are satisfied with smart watch, but not with smart band
 - Also, most smart watch user are satisfied with notification function but it is not healthcare function
 - Healthcare service and functions are required based on personal characteristics

• Factors of Long-term Engagement: Design, Durability, Battery (Usable time)

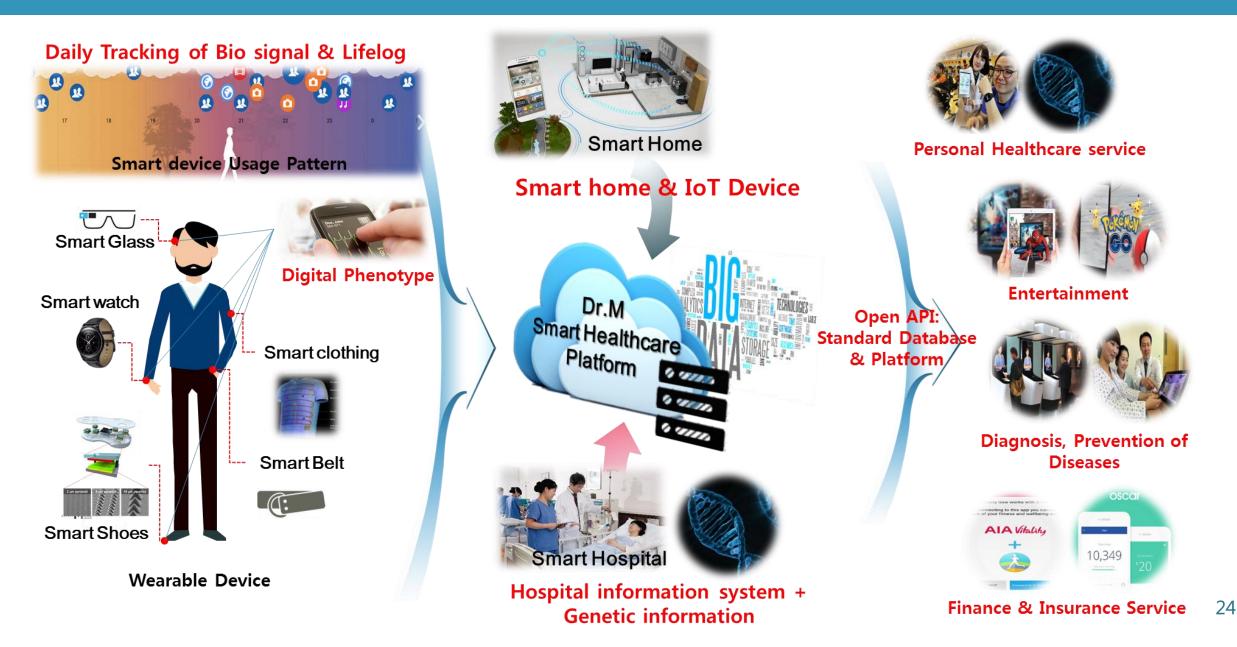
- Design ∝ wearing loyalty
- Frequent replacement and A/S \rightarrow stop using, negative recognition about smart device
 - At least 3months to 12 months, Most smart bands suffer damage to strap or broken down
- Frequent battery charging, charging station → stop wearing due to frequent loss, Spatial constraints of charging



"Dr.M" project Consortium



Future Plans



Thanks for attention!!

Contact Us & More Information https://www.facebook.com/dr.m.kaist dr.m@kaist.ac.kr