

# **Modeling and Discovering Human Behavior from Smartphone Sensing Life-Log Data**

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# Outline

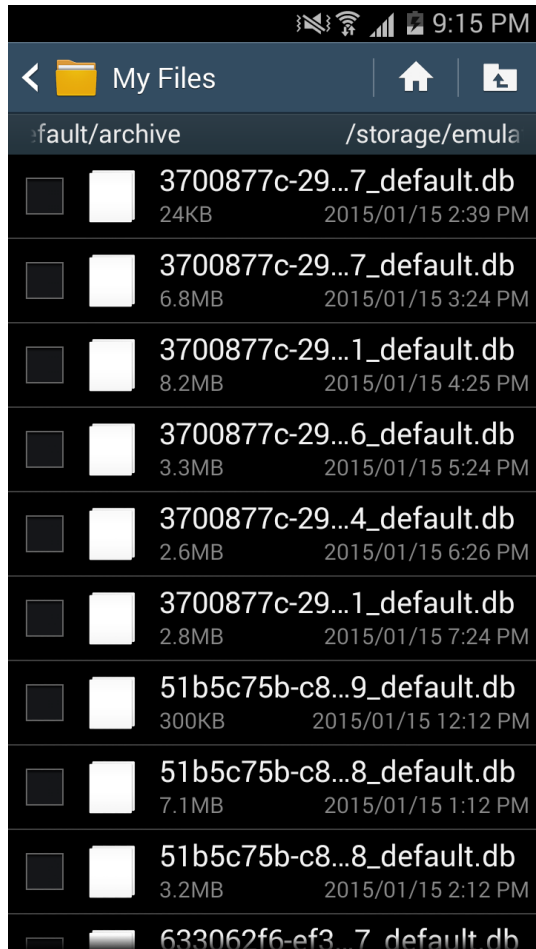
- **Introduction**
  - Background and Problem Statements
- **Methods**
  - Data Preprocessing and Features Extraction
  - Our approach details
- **Goal**
  - Discovering user behavior based on their smartphone life-log
  - Modeling human behavior for user identification

# Thesis Background and Problem Statements

- Common approach -> one features for one purpose.
- User has different types and brand of smartphone.
- We could not expect the human actions and activities.
- There is no ideal data collection which running for 24 hour non-stop.
- There is no ideal data collection that can record all of data without any data loss.

We decide to use many of sensors rather than focus only one sensor. Our proposed method tried to deal with those situations.

# Raw Data looks like



output — bash — 98x42

```
users-MacBook-Pro:output users$ ls
ENFP_0719      ESFP_0912      ESTP_4301      INTP_9712      ISTJ_2068      ISTJ_9139
ENFP_2012      ESFP_4634      ESTP_5154      ISFJ_2057      ISTJ_2837      ISTJ_9576
ENTJ_6454      ESTJ_3022      INFP_1993      ISFJ_2711      ISTJ_3052      ISTP_3948
ENTJ_6966      ESTJ_5071      INTJ_5498      ISFJ_7328      ISTJ_4667      ISTP_7676
ENTP_5623      ESTJ_5190      INTJ_7906      ISFP_4282      ISTJ_4700      XXXX_XXXX
ESFJ_2301      ESTJ_5824      INTP_3739      ISTJ_0178      ISTJ_4753      empty_error.txt
ESFJ_9284      ESTJ_6510      INTP_6399      ISTJ_0386      ISTJ_4968
```

users-MacBook-Pro:output users\$

_id	name	timestamp	value
Filter	Filter	Filter	Filter
1	HardwareInfoProbe	1404116791.153	{"androidId":"df03d9eae60fecb3","bluetoothMac":"88:9...
2	SmsProbe	1403883215.0	{"address":"15444302","body":{"ONE_WAY_HASH":"...
3	SmsProbe	1403883087.0	{"address":"15444302","body":{"ONE_WAY_HASH":"...
4	CallLogProbe	1404112293.842	{"_id":4108,"date":1404112293842,"duration":29,"name":...
5	SmsProbe	1403772474.744	{"address":"01066296348","body":{"ONE_WAY_HASH":"...
6	SmsProbe	1403697660.0	{"address":"0220338500","body":{"ONE_WAY_HASH":"...
7	CallLogProbe	1404111655.487	{"_id":4107,"date":1404111655487,"duration":75,"name":...
8	SmsProbe	1403697242.0	{"address":"0625101111","body":{"ONE_WAY_HASH":"...
9	CallLogProbe	1404111605.734	{"_id":4106,"date":1404111605734,"duration":0,"name":...
10	SmsProbe	1403609220.0	{"address":"0622395000","body":{"ONE_WAY_HASH":"...
11	SmsProbe	1403540566.46	{"address":"01029738808","body":{"ONE_WAY_HASH":"...
12	CallLogProbe	1404111295.005	{"_id":4105,"date":1404111295005,"duration":0,"name":...
13	SmsProbe	1403540204.981	{"address":"01086198051","body":{"ONE_WAY_HASH":"...

# Human Behavior

- Alice is research's student in one of university in Korea.
- Almost every day, he wakes up, takes a shower, breakfast, and goes to his campus at 8:40 AM.
- He is living in dormitory, he walks from dormitory to his lab (campus) takes 10 minutes.
- Usually, he arrives in his lab at 9 AM and then sits on his chair and starts working.

# Human behavior in case of smartphone sensing

- What is the human behavior in case of smartphone sensing?.
  - Human daily activities which carried out continuously
- In terms of human daily activities, we have to consider about four things:
  - What kind of activity (e.g meeting, studying, exercising)
  - When (e.g around 9 AM)
  - Location (e.g Lab)
  - Human Interaction (e.g all lab's members)
- Possibilities : same activity in different time and location, different activity in same time and location, etc.

# Dataset & Features Description

Proposed Features (*every single data has timestamp*)

- What kind of Human Activity
  - Activity [none, low, high]
- Human Location
  - GPS [longitude, latitude]
  - Bluetooth [list of nearby Bluetooth]
  - Wi-Fi [lists of nearby AP]
- Human Interaction (Human->Human)
  - Call [incoming, outgoing, missed]
  - SMS [sent, received]
  - Run apps [social network apps]
- Human Interaction (Human -> Smartphone)
  - Battery [time charging]
  - Run apps [name of apps]
  - Screen [screen ON, screen OFF]

**Table 2-2.** List of probes and types

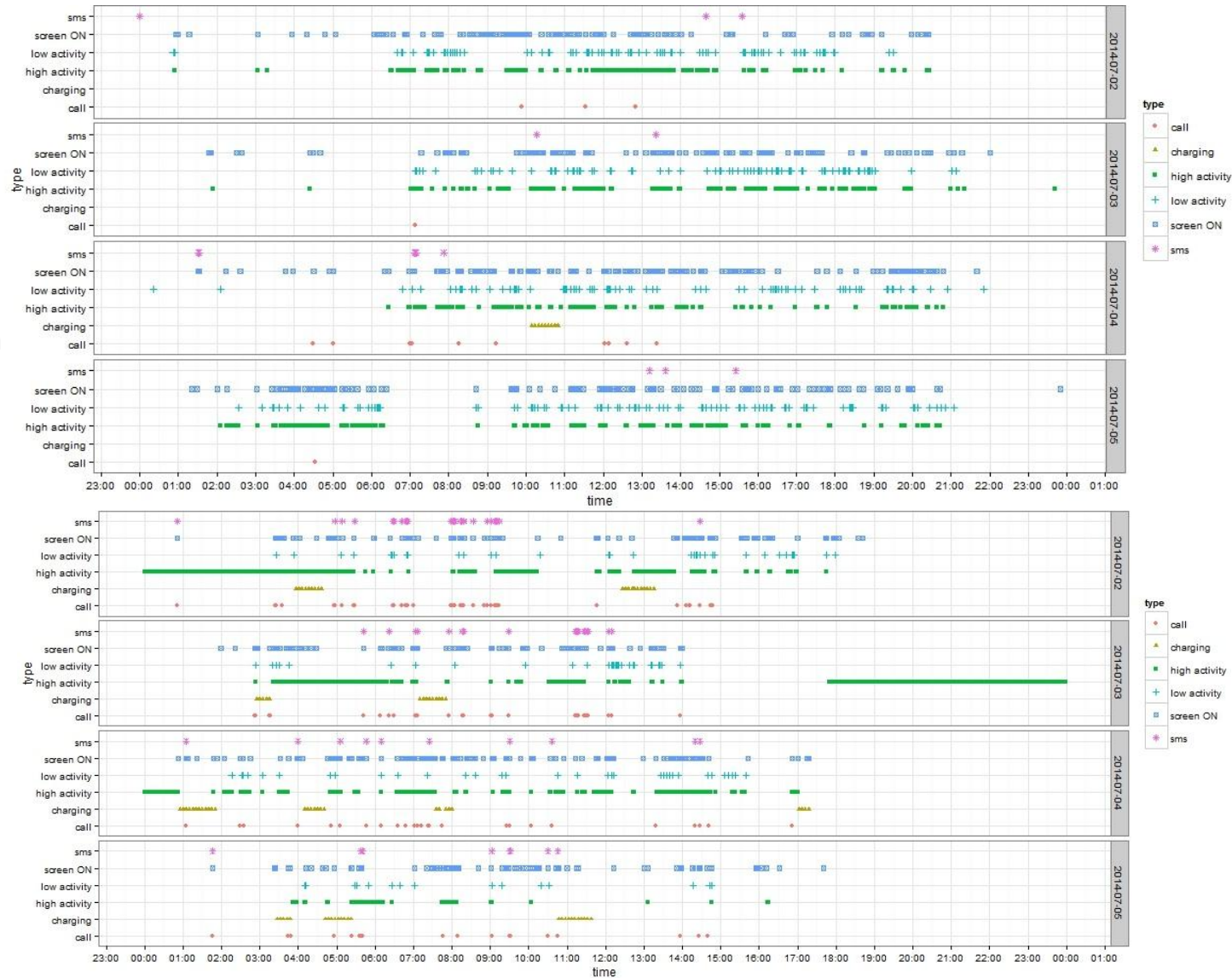
No.	Name of Probes	Explanation	Used
<b>On Request Data</b>			
1.	SimpleLocationProbe	GPS data (user location)	X
2.	WifiProbe	Nearby Wi-Fi signals	X
3.	BluetoothProbe	Nearby Bluetooth signals	X
4.	BatteryProbe	Battery status	X
<b>Historical Data</b>			
1.	CallLogProbe	User call log	X
2.	SmsProbe	User SMS log	X
3.	ApplicationsProbe	List of application installed	
4.	HardwareInfoProbe	User's smartphone hardware info	
5.	BrowserBookmarksProbe	User Bookmarks	
6.	BrowserSearchesProbe	User Browser log	
7.	ContactProbe	User contact (phonebook)	
<b>Continuous Data</b>			
1.	LightSensorProbe	Measures the ambient light level (illumination) in lx	
2.	ProximitySensorProbe	Measures the proximity of an object in cm relative to the view screen of a device.	
3.	TemperatureSensorProbe	Measures the temperature of the device in degrees Celsius (°C).	
4.	MagneticFieldSensorProbe	Measures the ambient geomagnetic field (x, y, z) in $\mu T$	
5.	PressureSensorProbe	Measures the ambient air pressure in hPa or mbar.	
6.	ScreenProbe	Screen phone (on and off)	X
7.	RunningApplicationsProbe	List of running applications	X
8.	ActivityProbe	User activity log based on accelerometer sensor (none, low, and high activity)	X

# Technical Explanation

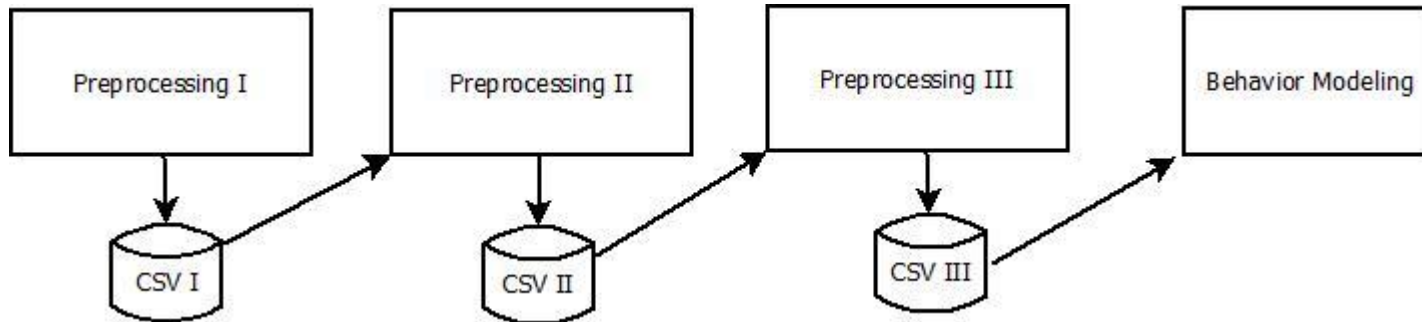


# Example data plot from two of students

1. We store the data from all of students in archive file.
2. The size of all of data after extracted is around **28.7 GB**.  
Extracted data contain 47 directories in different name for each student's data.
3. We used 37 students data.



# Pre-Processing Summarization



- Preprocessing I
  - Data cleansing
  - Removing duplication and noisy data
  - Select the most important data
- Preprocessing II
  - Features Extraction applied in here
- Preprocessing III
  - Features Aggregation
  - Adding new values based on features
  - Fitting the features data before modeling behavior applied

# Preprocessing I

```
output — bash — 98x42
users-MacBook-Pro:output user$ ls -al ENFP_0719/
total 68552
drwxr-xr-x  12 user  staff    408 Dec  3 12:06 .
drwxr-xr-x  43 user  staff   1462 Dec  3 15:20 ..
-rwxr-xr-x   1 user  staff  2872223 Dec  3 12:06 d_activity.csv
-rwxr-xr-x   1 user  staff  993357 Dec  3 12:06 d_battery.csv
-rwxr-xr-x   1 user  staff  211551 Dec  3 12:06 d_bluetooth.csv
-rwxr-xr-x   1 user  staff  7669244 Dec  3 12:06 d_call.csv
-rwxr-xr-x   1 user  staff   823183 Dec  3 12:06 d_location.csv
-rwxr-xr-x   1 user  staff  2437280 Dec  3 12:06 d_runapps.csv
-rwxr-xr-x   1 user  staff   635795 Dec  3 12:06 d_screen.csv
-rwxr-xr-x   1 user  staff    19 Dec  3 12:06 d_search.csv
-rwxr-xr-x   1 user  staff 13792370 Dec  3 12:06 d_sms.csv
-rwxr-xr-x   1 user  staff  5637443 Dec  3 12:06 d_wifi.csv
users-MacBook-Pro:output user$
```

1. *Funf* lib has problem in historical data such as SMS and call log.
2. We use 86400 second interval, means the application copy those data from android database system to our application database once every day.
3. It makes duplication in our database.
4. The function of this module are:
  - Removing duplication
  - Data cleansing
  - Select the most important data

# Preprocessing I output data looks like

ENFP\_0719 — more — 98x42

```
"" , "when", "latitude", "longitude"
"1", "2014-06-30 08:26:36", "35.1754255", "126.9130674"
"2", "2014-06-30 08:31:30", "35.1754281", "126.9130184"
"3", "2014-06-30 08:36:30", "35.1754225", "126.9130536"
"4", "2014-06-30 08:41:30", "35.1753982", "126.9130597"
"5", "2014-06-30 08:46:30", "35.1753991", "126.9130549"
"6", "2014-06-30 08:51:30", "35.1753997", "126.9130598"
"7", "2014-06-30 08:56:31", "35.1754333", "126.9130484"
"8", "2014-06-30 09:01:30", "35.1754143", "126.9130638"
"9", "2014-06-30 09:06:30", "35.1754126", "126.9130687"
"10", "2014-06-30 09:11:30", "35.1754142", "126.9130553"
"11", "2014-06-30 09:16:30", "35.1753987", "126.913074"
"12", "2014-06-30 09:21:30", "35.1754178", "126.9130686"
"13", "2014-06-30 09:26:30", "35.1754176", "126.9130779"
"14", "2014-06-30 09:31:30", "35.1753914", "126.9130496"
"15", "2014-06-30 09:36:30", "35.1754066", "126.9130196"
"16", "2014-06-30 09:41:30", "35.1753833", "126.9130507"
"17", "2014-06-30 09:46:30", "35.1753883", "126.9130474"
"18", "2014-06-30 09:51:52", "35.174465", "126.9135807"
"19", "2014-06-30 09:56:30", "35.174465", "126.9135807"
"20", "2014-06-30 10:01:30", "35.1735813", "126.9119189"
"21", "2014-06-30 10:06:53", "35.1744781", "126.9137426"
"22", "2014-06-30 10:11:30", "35.1735813", "126.9119189"
"23", "2014-06-30 10:16:52", "35.174502", "126.9137499"
"24", "2014-06-30 10:21:30", "35.1749091", "126.9124954"
"25", "2014-06-30 10:26:30", "35.1749091", "126.9124954"
"26", "2014-06-30 10:31:30", "35.1749091", "126.9124954"
```

ENFP\_0719 — more — 98x42

```
"" , "timestamp", "duration", "package"
"1", "2014-06-30 08:26:31", "12.401", "edu.mit.media.funf.wifiscanner"
"2", "2014-06-30 08:26:43", "2.055", "com.nhn.android.search"
"3", "2014-06-30 08:26:45", "9.183", "com.buzzpia.aqua.launcher"
"4", "2014-06-30 08:26:54", "15.32", "edu.mit.media.funf.wifiscanner"
"5", "2014-06-30 08:27:10", "38.126", "com.rechild.advancedtaskkiller"
"6", "2014-06-30 08:27:48", "3.015", "edu.mit.media.funf.wifiscanner"
"7", "2014-06-30 08:27:51", "6.015", "com.buzzpia.aqua.launcher"
"8", "2014-06-30 08:27:57", "2.005", "edu.mit.media.funf.wifiscanner"
"9", "2014-06-30 08:27:59", "2.006", "com.buzzpia.aqua.launcher"
"10", "2014-06-30 08:28:01", "2.044", "com.kakao.talk"
"11", "2014-06-30 08:28:03", "6.042", "com.buzzpia.aqua.launcher"
"12", "2014-06-30 08:28:09", "63.362", "com.nhn.android.search"
"13", "2014-06-30 08:29:12", "0.841", "com.buzzpia.aqua.launcher"
"14", "2014-06-30 08:30:02", "55.247", "com.buzzpia.aqua.launcher"
"15", "2014-06-30 08:33:35", "9.898", "com.buzzpia.aqua.launcher"
"16", "2014-06-30 08:33:46", "4.015", "com.buzzpia.aqua.launcher"
"17", "2014-06-30 08:33:50", "68.308", "com.kakao.talk"
"18", "2014-06-30 08:34:58", "9.017", "com.kakao.talk"
"19", "2014-06-30 08:35:07", "3.006", "com.buzzpia.aqua.launcher"
"20", "2014-06-30 08:35:10", "205.598", "com.facebook.katana"
"21", "2014-06-30 08:38:36", "1.245", "com.buzzpia.aqua.launcher"
"22", "2014-06-30 08:40:27", "5.011", "com.buzzpia.aqua.launcher"
"23", "2014-06-30 08:40:32", "61.606", "com.joeware.android.gpulumera"
"24", "2014-06-30 08:41:34", "13.806", "com.buzzpia.aqua.launcher"
```

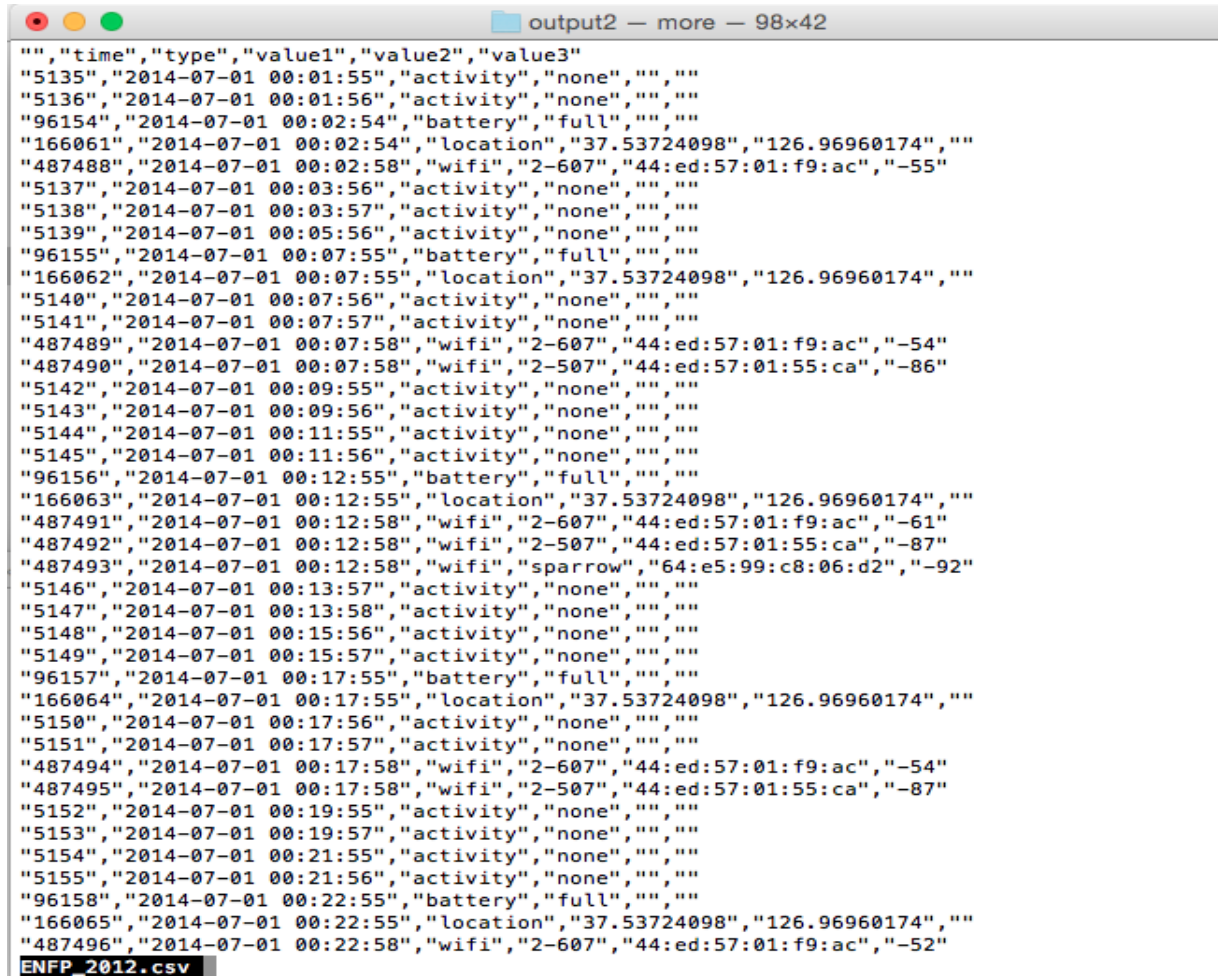


# Pre-processing II (Features Extraction)

**Table 2-4.** List of features and the values

No	Name of Probes	Value1	Value2	Value3
1.	ActivityProbe	Status ( <i>"none"</i> , <i>"low"</i> , and <i>"high"</i> )		
2.	SimpleLocationProbe	Latitude	Longitude	
3.	WifiProbe	List of nearby SSID	MAC	Signal strength (dB)
4.	BluetoothProbe	List of nearby Bluetooth devices		
5.	BatteryProbe	Status ( <i>"discharging"</i> , <i>"full"</i> , and <i>"charging"</i> )		
6.	ScreenProbe	ON/OFF		
7.	RunningApplicationsProbe	Apps name	Duration	
8.	CallLogProbe	Number	Types	Duration
9.	SmsProbe	Number	Types	Text length

# Preprocessing II output data looks like



```
output2 - more - 98x42
", "time", "type", "value1", "value2", "value3"
"5135", "2014-07-01 00:01:55", "activity", "none", "", ""
"5136", "2014-07-01 00:01:56", "activity", "none", "", ""
"96154", "2014-07-01 00:02:54", "battery", "full", "", ""
"166061", "2014-07-01 00:02:54", "location", "37.53724098", "126.96960174", ""
"487488", "2014-07-01 00:02:58", "wifi", "2-607", "44:ed:57:01:f9:ac", "-55"
"5137", "2014-07-01 00:03:56", "activity", "none", "", ""
"5138", "2014-07-01 00:03:57", "activity", "none", "", ""
"5139", "2014-07-01 00:05:56", "activity", "none", "", ""
"96155", "2014-07-01 00:07:55", "battery", "full", "", ""
"166062", "2014-07-01 00:07:55", "location", "37.53724098", "126.96960174", ""
"5140", "2014-07-01 00:07:56", "activity", "none", "", ""
"5141", "2014-07-01 00:07:57", "activity", "none", "", ""
"487489", "2014-07-01 00:07:58", "wifi", "2-607", "44:ed:57:01:f9:ac", "-54"
"487490", "2014-07-01 00:07:58", "wifi", "2-507", "44:ed:57:01:55:ca", "-86"
"5142", "2014-07-01 00:09:55", "activity", "none", "", ""
"5143", "2014-07-01 00:09:56", "activity", "none", "", ""
"5144", "2014-07-01 00:11:55", "activity", "none", "", ""
"5145", "2014-07-01 00:11:56", "activity", "none", "", ""
"96156", "2014-07-01 00:12:55", "battery", "full", "", ""
"166063", "2014-07-01 00:12:55", "location", "37.53724098", "126.96960174", ""
"487491", "2014-07-01 00:12:58", "wifi", "2-607", "44:ed:57:01:f9:ac", "-61"
"487492", "2014-07-01 00:12:58", "wifi", "2-507", "44:ed:57:01:55:ca", "-87"
"487493", "2014-07-01 00:12:58", "wifi", "sparrow", "64:e5:99:c8:06:d2", "-92"
"5146", "2014-07-01 00:13:57", "activity", "none", "", ""
"5147", "2014-07-01 00:13:58", "activity", "none", "", ""
"5148", "2014-07-01 00:15:56", "activity", "none", "", ""
"5149", "2014-07-01 00:15:57", "activity", "none", "", ""
"96157", "2014-07-01 00:17:55", "battery", "full", "", ""
"166064", "2014-07-01 00:17:55", "location", "37.53724098", "126.96960174", ""
"5150", "2014-07-01 00:17:56", "activity", "none", "", ""
"5151", "2014-07-01 00:17:57", "activity", "none", "", ""
"487494", "2014-07-01 00:17:58", "wifi", "2-607", "44:ed:57:01:f9:ac", "-54"
"487495", "2014-07-01 00:17:58", "wifi", "2-507", "44:ed:57:01:55:ca", "-87"
"5152", "2014-07-01 00:19:55", "activity", "none", "", ""
"5153", "2014-07-01 00:19:57", "activity", "none", "", ""
"5154", "2014-07-01 00:21:55", "activity", "none", "", ""
"5155", "2014-07-01 00:21:56", "activity", "none", "", ""
"96158", "2014-07-01 00:22:55", "battery", "full", "", ""
"166065", "2014-07-01 00:22:55", "location", "37.53724098", "126.96960174", ""
"487496", "2014-07-01 00:22:58", "wifi", "2-607", "44:ed:57:01:f9:ac", "-52"
ENFP_2012.csv
```

## Preprocessing III

- Temporal Granularity (round time value)
  - < :30 -> round down
  - > :30 -> round up
- Changing Location value to (*“same”, “little”, “long”*)
  - 0.0001 degree = 11.1132 m (*“little”*: “between 0.001 ~ 0.005”)
- Aggregate values of Wi-Fi and Bluetooth
- Removing values such as text length and duration from SMS log and call log, duration from running application probe, MAC and signal strength from nearby Wi-Fi probe.

# Output of Preprocessing III

```
output3 — more — 108x42
Timestamp,Weekday,HP,Sensor Name,Sensor Value
07/01/2014 0:01,Tuesday,00:00,location,long
07/01/2014 0:01,Tuesday,00:00,wifi,KT_WLAN_C1BF, iptime
07/01/2014 0:02,Tuesday,00:00,activity,low
07/01/2014 0:02,Tuesday,00:00,activity,low
07/01/2014 0:04,Tuesday,00:00,runapps,com.buzzpia.aqua.launcher
07/01/2014 0:04,Tuesday,00:00,runapps,com.facebook.katana
07/01/2014 0:04,Tuesday,00:00,runapps,com.buzzpia.aqua.launcher
07/01/2014 0:04,Tuesday,00:00,runapps,net.daum.android.cafe
07/01/2014 0:06,Tuesday,00:00,location,long
07/01/2014 0:06,Tuesday,00:00,wifi,KT_WLAN_C1BF, iptime
07/01/2014 0:11,Tuesday,00:00,location,long
07/01/2014 0:11,Tuesday,00:00,wifi,KT_WLAN_C1BF, iptime
07/01/2014 0:14,Tuesday,00:00,activity,low
07/01/2014 0:14,Tuesday,00:00,activity,low
07/01/2014 0:15,Tuesday,00:00,runapps,com.buzzpia.aqua.launcher
07/01/2014 0:15,Tuesday,00:00,runapps,com.nhn.android.search
07/01/2014 0:16,Tuesday,00:00,runapps,net.daum.android.cafe
07/01/2014 0:16,Tuesday,00:00,location,long
07/01/2014 0:16,Tuesday,00:00,wifi,KT_WLAN_C1BF, iptime
07/01/2014 0:19,Tuesday,00:00,runapps,com.nhn.android.search
07/01/2014 0:20,Tuesday,00:00,runapps,net.daum.android.cafe
07/01/2014 0:21,Tuesday,00:00,location,long
07/01/2014 0:21,Tuesday,00:00,wifi,KT_WLAN_C1BF
07/01/2014 0:26,Tuesday,00:00,location,long
07/01/2014 0:26,Tuesday,00:00,wifi,KT_WLAN_C1BF
07/01/2014 0:31,Tuesday,01:00,location,long
07/01/2014 0:31,Tuesday,01:00,wifi,KT_WLAN_C1BF, iptime
07/01/2014 0:36,Tuesday,01:00,activity,high
07/01/2014 0:36,Tuesday,01:00,location,long
07/01/2014 0:36,Tuesday,01:00,activity,high
07/01/2014 0:36,Tuesday,01:00,wifi,KT_WLAN_C1BF
07/01/2014 0:37,Tuesday,01:00,runapps,com.facebook.katana
07/01/2014 0:37,Tuesday,01:00,runapps,net.daum.android.cafe
07/01/2014 0:40,Tuesday,01:00,activity,high
07/01/2014 0:41,Tuesday,01:00,location,long
07/01/2014 0:41,Tuesday,01:00,wifi,KT_WLAN_C1BF, iptime
07/01/2014 0:46,Tuesday,01:00,location,long
07/01/2014 0:46,Tuesday,01:00,activity,low
07/01/2014 0:46,Tuesday,01:00,wifi,KT_WLAN_C1BF
07/01/2014 0:51,Tuesday,01:00,location,long
INTJ_5498_HP.csv
```



# Discovering Human Behavior

1. The data that we have are set of activities.
2. Behavior means activities which is carried out continuously.
3. Behavior means set of group activities which has (\* similar time, location and similar activity in different days).

Day1 Week1	Day2 Week1	Day3 Week1	Day4 Week1	Day5 Week1	Day6 Week1	<del>Day7 Week1</del>	Day1 Week2
---------------	---------------	---------------	---------------	---------------	---------------	---------------------------	---------------

# Finding Similar Patterns

Time	Sensor Name	Sensor Value
13:00	location	same
13:00	wifi	1-AP, iptime
14:00	runapps	kakao
14:00	location	long
15:00	runapps	kakao
15:00	location	little

Time	Sensor Name	Sensor Value
13:00	location	same
13:00	wifi	1-AP, iptime
14:00	battery	charging
14:00	wifi	D-link
15:00	runapps	kakao
15:00	location	little

Group-1 = 13:00,location,same | 13:00,wifi,1-AP,iptime

Group-1 = 13:00,location,same | 13:00,wifi,1-AP,iptime

Group-2 = 15:00,runapps,kakao | 15:00, location, little

Group-2 = 15:00,runapps,kakao | 15:00, location, little

# Algorithm (Similarity Detection)

**Data** : D, w

**Result** : All Detected Group in a Window

```
grpAll, grpTemp, grpPrevious<- NULL
```

```
dataValue, dataValueNext <- NULL
```

```
while (D in w) for all of D do
```

```
    dataValue <- D.current.day
```

```
    dataValueNext <- D.next.day
```

```
    grpTemp <- findingSimilarPatterns(dataValue, dataValueNext)
```

```
    if (grpTemp in grpPrevious) then
```

```
        grpNew <- merge(grpPrevious, grpTemp)
```

```
        grpAll <- add(grpNew)
```

```
    else
```

```
        grpAll <- add(grpTemp)
```

## Behavior Profiling/Modeling

We collect all of intersection data between Groups, and mark those data as the user behaviors.

# Grouping Result

```
G2,"19:00,location,same|19:00,bluetooth,DTVBluetooth|"
G2,"19:00,location,same|19:00,bluetooth,DTVBluetooth|"
G2,"19:00,location,same|19:00,bluetooth,DTVBluetooth|"
G2,"19:00,location,same|19:00,bluetooth,DTVBluetooth|"
G2,"19:00,location,same|19:00,bluetooth,DTVBluetooth|"
G3,"20:00,location,same|20:00,bluetooth,DTVBluetooth|"
G3,"20:00,location,same|20:00,bluetooth,DTVBluetooth|"
G3,"20:00,location,same|20:00,bluetooth,DTVBluetooth|"
G3,"20:00,location,same|20:00,bluetooth,DTVBluetooth|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G7,"06:00,screen,OFF|06:00,location,same|"
G9,"12:00,battery,charging|12:00,bluetooth,ESVH-PC|12:00,location,same|12:00,runapps,com.lge.launcher2|"
G9,"12:00,battery,charging|12:00,bluetooth,ESVH-PC|12:00,location,same|12:00,runapps,com.lge.launcher2|"
G9,"12:00,battery,charging|12:00,bluetooth,ESVH-PC|12:00,location,same|12:00,runapps,com.lge.launcher2|"
G9,"12:00,battery,charging|12:00,bluetooth,ESVH-PC|12:00,location,same|12:00,runapps,com.lge.launcher2|"
G9,"12:00,battery,charging|12:00,bluetooth,ESVH-PC|12:00,location,same|12:00,runapps,com.lge.launcher2|"
G9,"12:00,battery,charging|12:00,bluetooth,ESVH-PC|12:00,location,same|12:00,runapps,com.lge.launcher2|"
G9,"12:00,battery,charging|12:00,bluetooth,ESVH-PC|12:00,location,same|12:00,runapps,com.lge.launcher2|"
INTJ_8928_groups.csv
```

# Testing for Human Identification

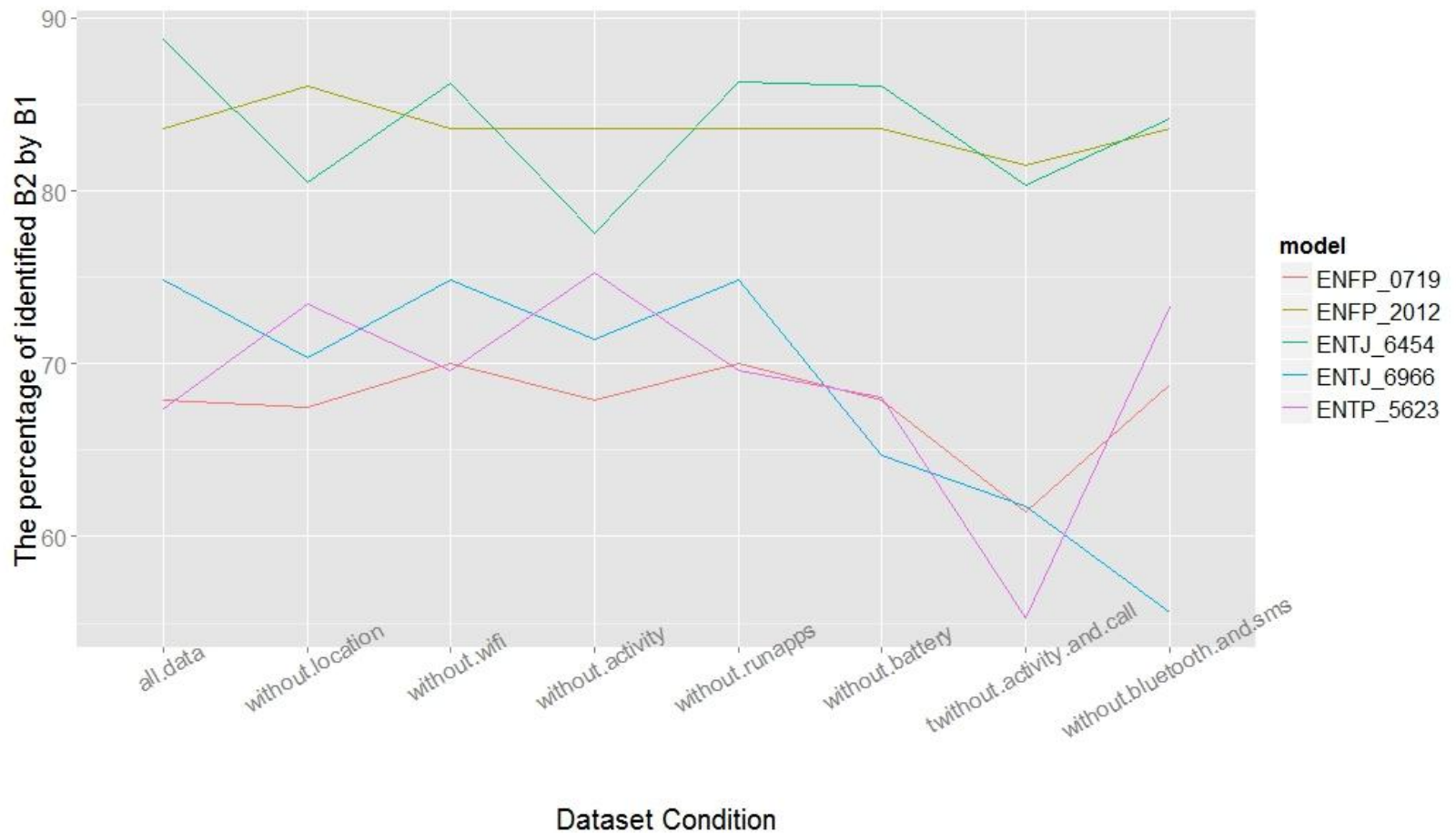
# Performance Evaluation

- Total of dataset around 1 month 20 days
- We Divide all of dataset to two parts
  - First month for creating model (first dataset)
  - Remaining dataset for testing performance (second dataset)
- Modeling user behavior based on first data,
  - B1: Behavior model/profile.
- Extract and Process the second dataset.
- Apply similarity detection to second dataset with same setting.
  - B2: Set of behavior group result from second dataset.
- Is the all of B2 identified by Behavior model/profile (B1)?.
- How many groups of activities (B2) which identified by behavior model(B1), The percentage of data identified.
- Implement this approach for Identification

# Identification result (Only 6 students)

	TEST						
		ENFP_0719	ENFP_2012	INTJ_5498	ISTJ_3052	ESTJ_5190	ESFP_4634
MODEL	ENFP_0719	67.922	0	0.4	2.187	0	1.943
	ENFP_2012	0	83.582	0	0	0	0
	INTJ_5498	2.178	0	75.977	2.087	0	3.401
	ISTJ_3052	2.289	0	0.4	93.439	8.232	1.943
	ESTJ_5190	0	0	0	0.099	22.866	0
	ESFP_4634	2.289	0	0.977	2.087	0	89.686

# Testing Performance by Removing Some Features





# Conclusion and Future Works

- **Conclusion**

- In this thesis, we proposed approach that can used for user identification by building human behavior model based on smartphone life-log data.
- We use and combine of many sensors instead only focus on one sensors because we realize that sometimes user does not has data from one or more sensors.
- Based on our result, we can see that our approach is good enough for user identification. We have tried also to remove one or more features and then observe the accuracy values.

- **Future Works**

- Change the size of window
- Use different time precision
- Use size window in vertical when compare between days