

Identifying Household Activities from Smart Meter Data


ECEEE, June 2015

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Smart meters tell us about real-time energy use, but they don't tell us **why**, nor **what to do about it**.

(1) The routines and scheduling of domestic life explain why and when energy is used.

(2) To be meaningful and salient, energy feedback needs to reflect households' lived experience.

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activities: *what did you do today?*



Activities are described by time use.

Collecting data on time use is effortful and costly.



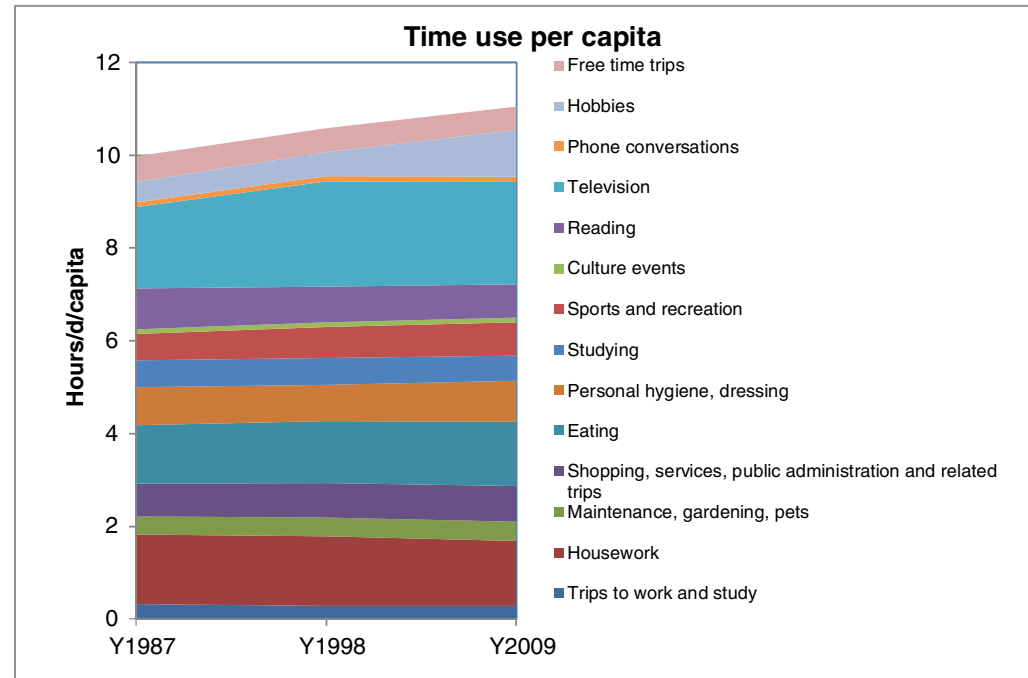
The Time Use Survey, 2005

How we spend our time

A report on research using the ONS Omnibus Survey produced on behalf of the Economic and Social Research Council (ESRC), Department of Culture, Media and Sport (DCMS), Department for Education and Skills (DfES), Department of Health (DH), Department for Transport (DfT), Office for National Statistics (ONS)

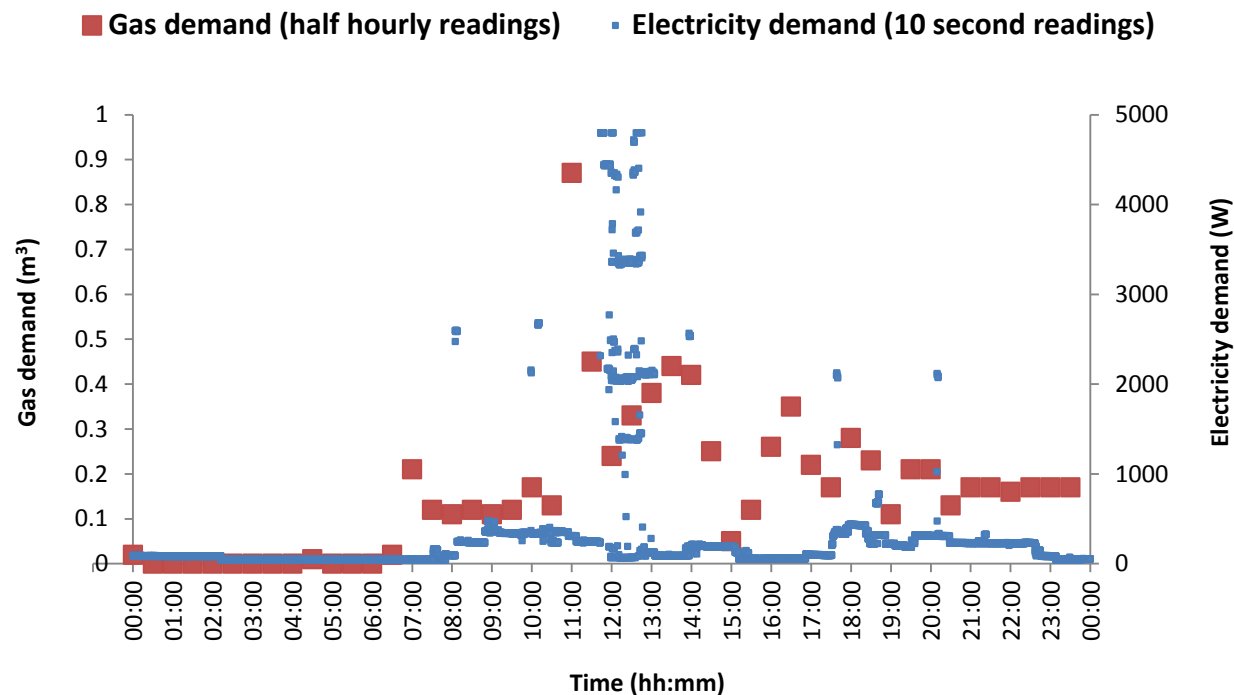
Deborah Lader
Sandra Short
Jonathan Gershuny
July 2006

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London SW1V 2QQ
Tel: 020 7533 9233
Fax: 020 7533 9292



Jalas & Juntunen (2015). *Ecological Economics* 113: 51-59.

Can we make **reliable inferences** about household **activities** using smart meter data?



Firth, S. (2014). "The REFIT Household Study." TEDDInet conference. April 2014.

Energy Saving Advice Service 0300 123 1234 HELPING CONSUMERS | 19

Smart meters

What is a smart meter?
Smart meters are the next generation of gas and electricity meters and provide near real time information about the amount of energy a household uses. Smart meters will bring an end to estimated billing.
As well as the meter itself, householders will also be offered an In Home Display which shows how much energy is being used in pounds and pence, and should help households to budget better.
Together, the In Home Display linked to a smart meter will put households in control, and help them to avoid wasting money and energy.

When will households get smart meters?
Most households will have smart meters installed between 2015 and 2020, although some energy companies are starting to install smart meters now.

How much will it cost to get a smart meter installed?
Households will not be charged separately for a smart meter or for the In Home Display.

Consumer protection
The Government is ensuring that consumer benefits and protection are at the heart of the smart meter programme.

This means:

- No sales during the installation visit
- Installers must provide energy efficiency advice as part of the visit and they will need the consumer's permission before the visit if they want to talk to them about their own products
- Protecting the privacy of individuals and putting them in control of smart meter data

Suppliers and consumer groups are working together to ensure consumers get accurate and helpful information about their smart meter.

By the end of June 2013 almost 90,000 smart meters have been installed in homes across Great Britain

Little things save a lot
Lynne is a busy stay-at-home mum with 2 young children; Daniel, aged 7 and Connor, aged 2. The family home is a 3 bedroom semi-detached property built in the 1970s.
Energy cost saving achieved
£76.49
Through a series of small daily changes, Lynne has managed to reduce her energy use by 14% in just 12 months.

"My smart meter really helped me see what appliances used most energy in our home. Not only is the smart meter saving us money, it's also helped us adopt some great energy saving habits which we are passing on to our children"

NAVIGATING DEBTS ENERGY EFFICIENCY ADVICE MORE CHOICES FURTHER INFORMATION

Method ... Results ... Applications

Method ... results ... applications

Method (6 Steps)

1. Define activities
2. Collect energy and household data
3. Disaggregate electricity data
4. Map technology <-> activity relationships
5. Make activity inferences
6. Validate activity inferences

Method: *Step 1*. Develop a comprehensive set of **activities** to describe household life and energy use.

DAILY ROUTINES:



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DAILY ROUTINES:

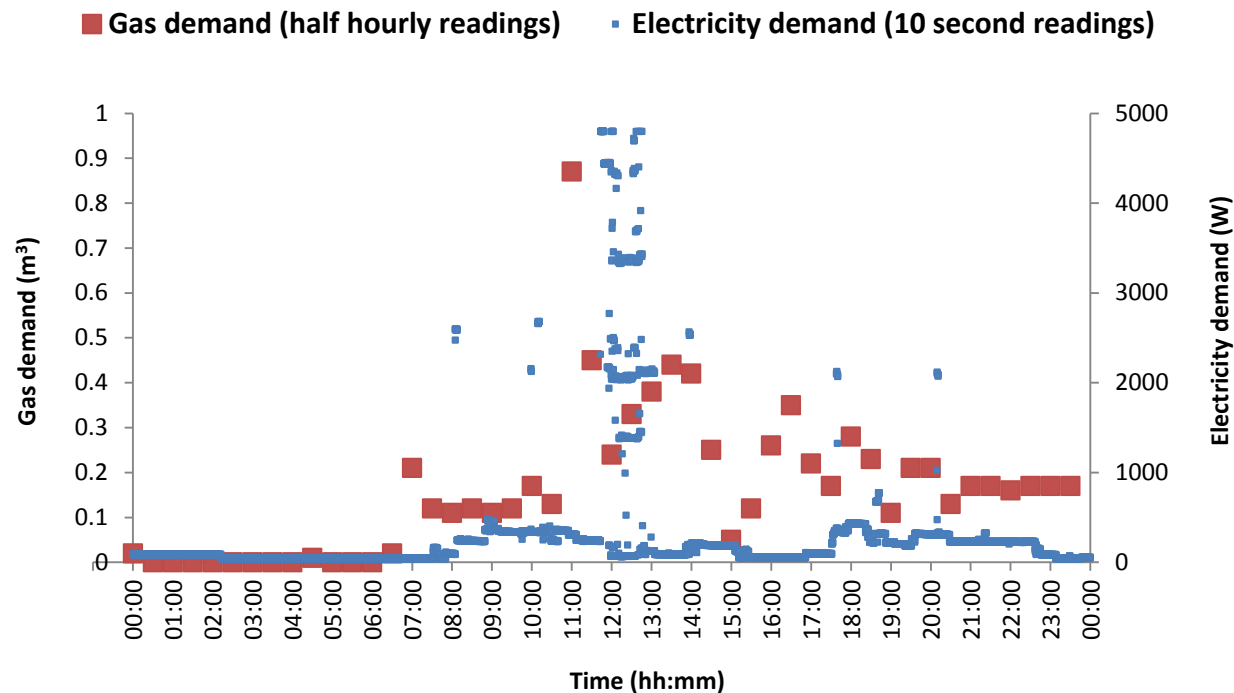


INTERACTING:
communicating
socialising

LEISURE & ICTs:
watching TV
listening to radio/music
gaming
computing

OTHER ACTIVITIES:
hobbies
caring
working
other

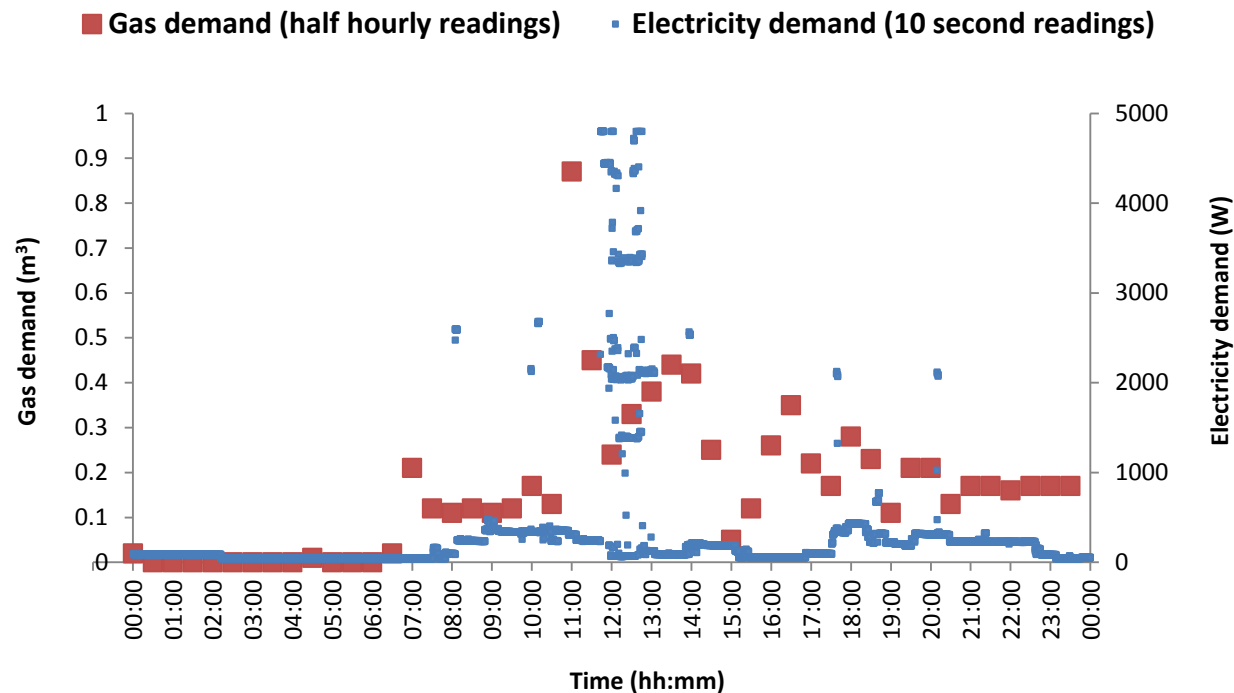
Method: *Step 2*. Collect **data** on **energy** & household routines from 6 homes (of 20 homes in a field trial).



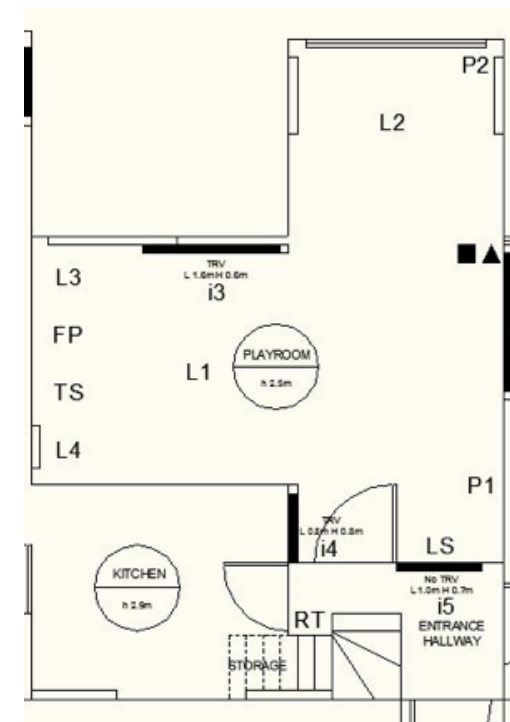
Firth, S. (2014). "The REFIT Household Study." TEDDInet conference. April 2014.



Method: *Step 2*. Collect **data** on energy & household routines from 6 homes (of 20 homes in a field trial).



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Method: *Step 3. Disaggregate* real-time electricity data.

- appliance-specific plug monitors
- non-intrusive appliance load monitoring (NALM)
 - validated by appliance time diaries

Method: *Step 4*. Map **relationships** between activities and technologies in an ‘activities ontology’.

APPLIANCES		ACTIVITIES									
		daily routines				ICT-related leisure					
Type	Appliance / Technology	cooking	eating	washing	laundrying	cleaning	sleeping	tv	radio	games	computing
electricity	kettle	x	o								
electricity	toaster	x	o								
electricity	microwave	x	o								
elecricity	washing machine				x						
electricity	vacuum cleaner					x					
electricity	food processor	x									
electricity	PC									x	x
electricity	laptop									x	x

+ additional information on:
 - location
 - typical usage patterns

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electricity	vacuum cleaner					x					
electricity	food processor	x									
electricity	PC									x	x
electricity	laptop									x	x

+ additional information on:
 - location
 - typical usage patterns

x = marker technology
~ = auxiliary technology
o = associated activity

Method ... results ... applications

Method (6 Steps)

1. Define activities
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6. Validate activity inferences

Method ... **Results** ... applications

Key Results (so far)

1. 6-8 activities are inferrable per household
2. Some time use profiles are stable, others vary
 - within household
 - between household
3. Some activities are temporally inter-linked, others are not
4. Activities usefully and meaningfully characterise energy use

*Results 1. Activity-inference methodology applied to
6 homes for 1 month (October 2014).*

House	Size & Composition
House 4	2 retired adults
House 8	2 retired adults
House 2	2 adults + 2 pre-school children
House 10	2 adults + 2 pre-school children
House 5	2 adults + 2 school children
House 19	2 adults + 2 school children

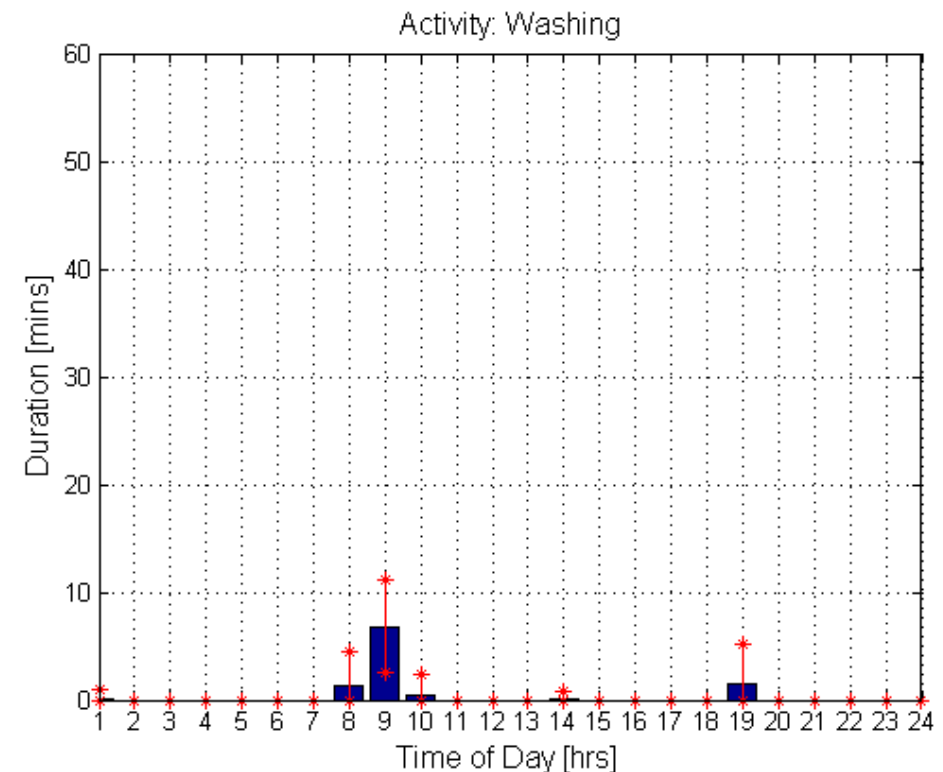
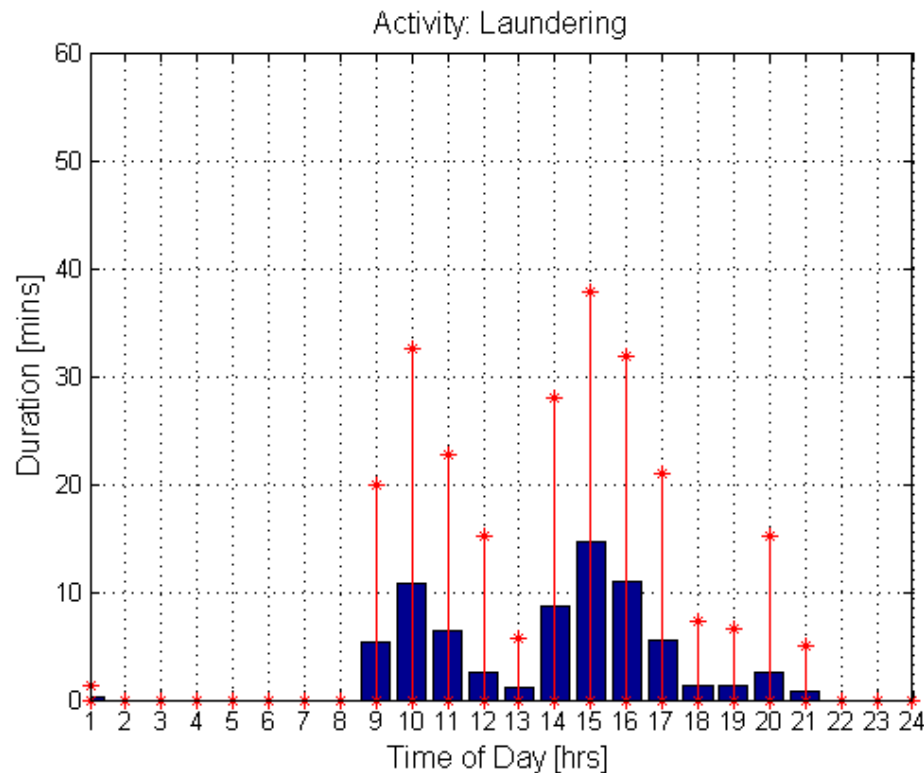
Results 1. 6-8 activities were **inferred** (remainder were not inferable or **not occurring** in that home).

	daily routines						interacting	ICT-related leisure				other				
House	cooking	eating	washing	laundrying	cleaning	sleeping	communicating	socialising	watching tv	listening to radio	gaming	computing	hobbies	caring	working	other
House 4																
House 8																
House 2																
House 10																
House 5																
House 19																



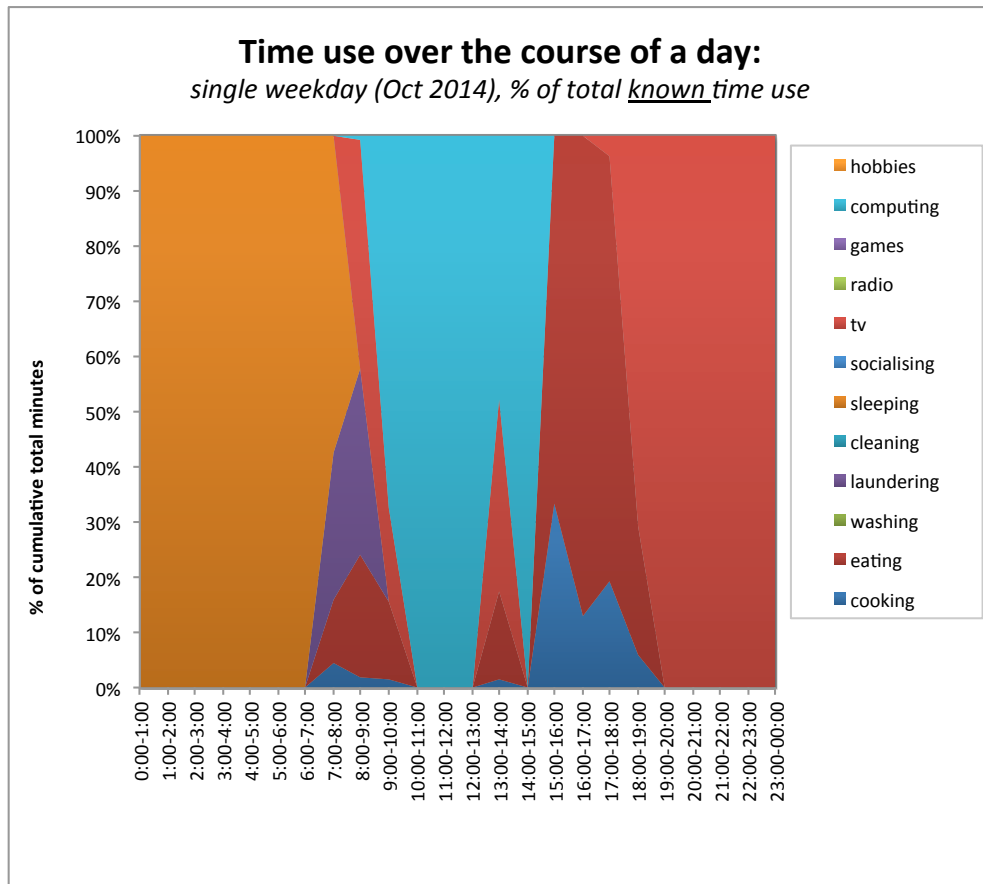
Results 2. Activities varied in the **regularity** of their **routine** (when and for how long).

mean (and 90% C.I.) of hourly time use profiles over a month



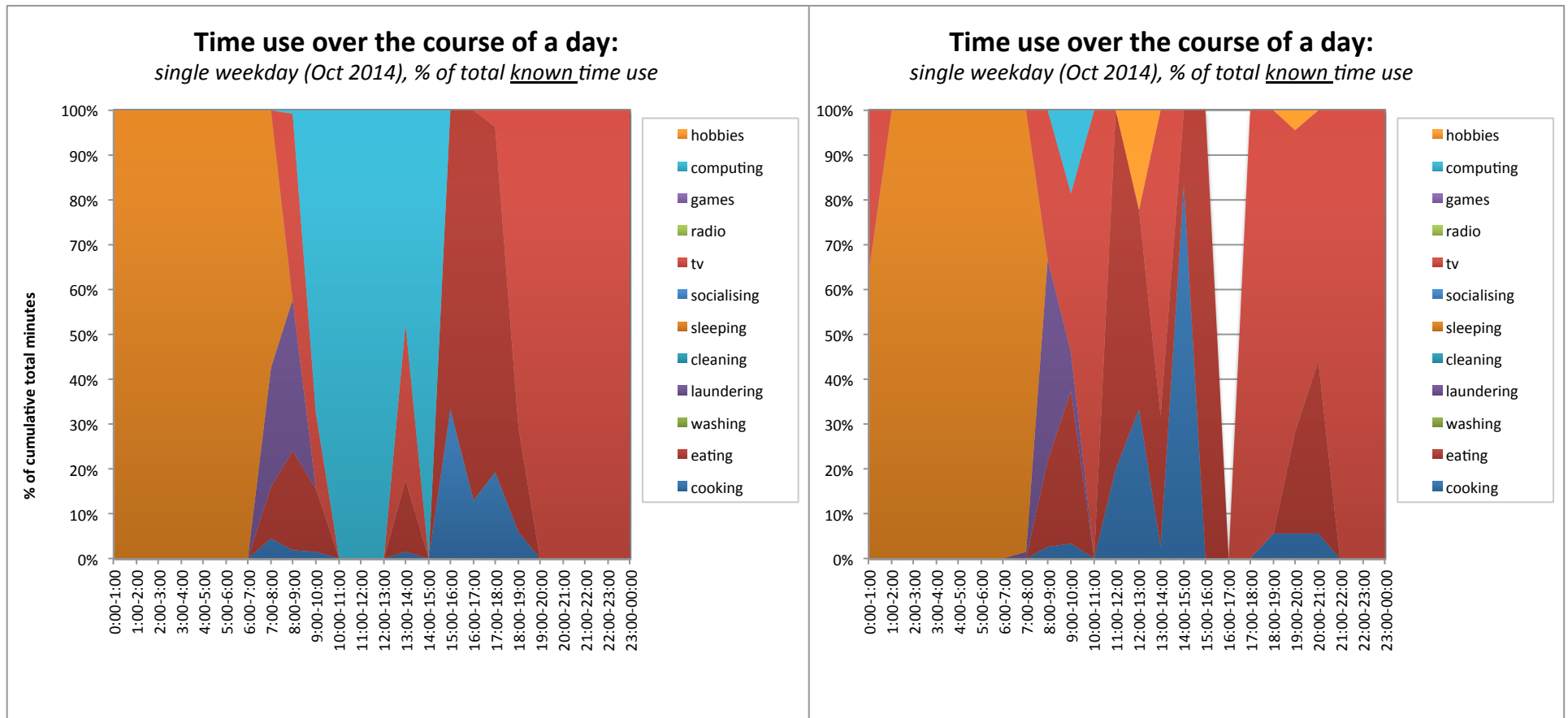
House 2. 2 adults + 2 pre-school children

Results 2. Some ... but not all activity time profiles varied from day to day.



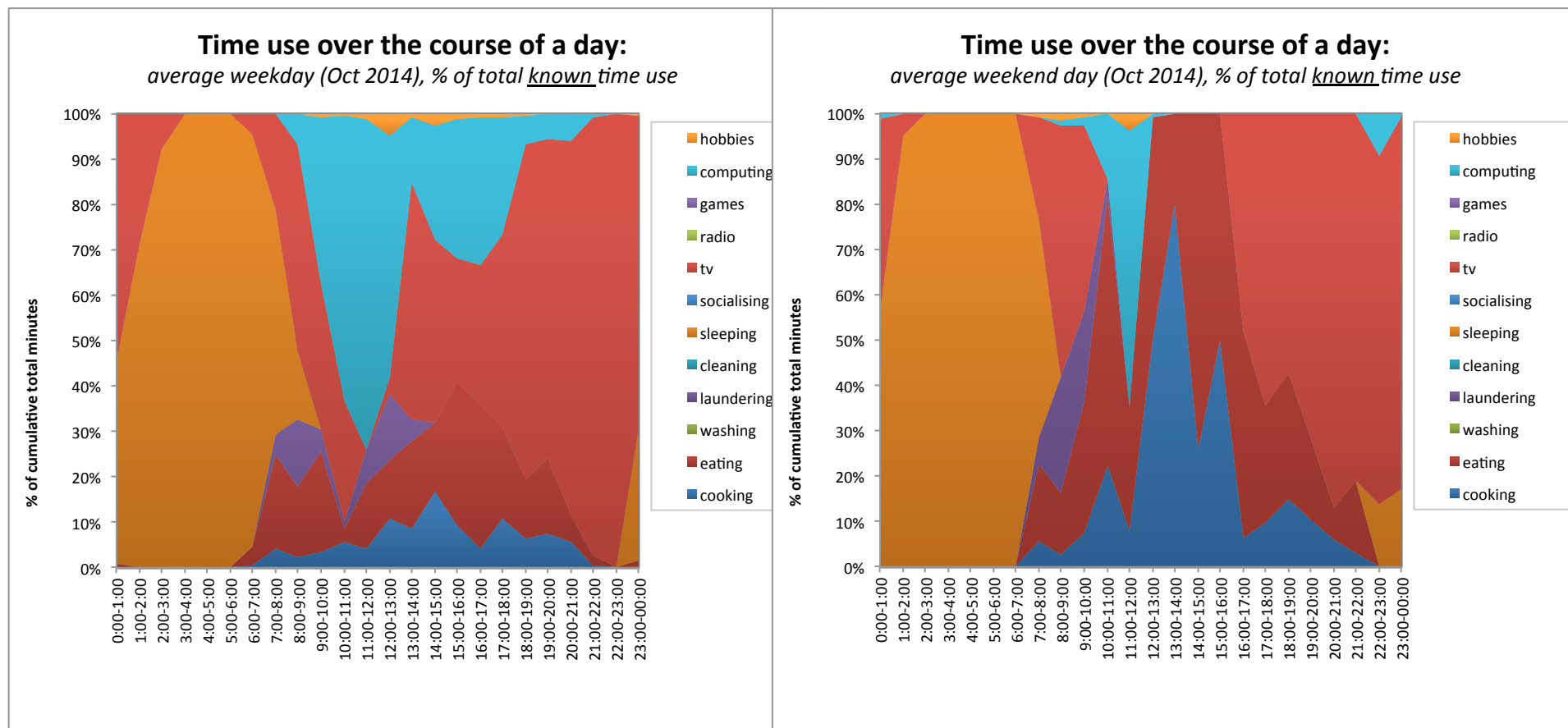
House 4. 2 retired adults

Results 2. Some ... but not all activity time profiles varied from day to day.



House 4. 2 retired adults

Results 2. Some ... but not all activity time profiles varied markedly between weekdays and weekends.

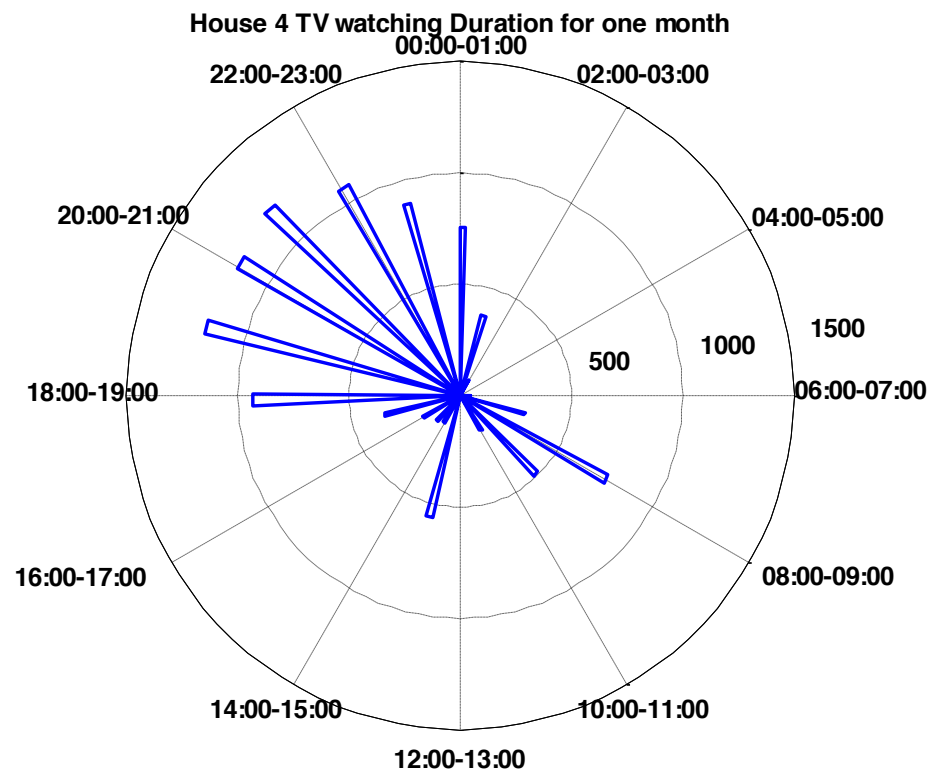


* computing *

* cooking *

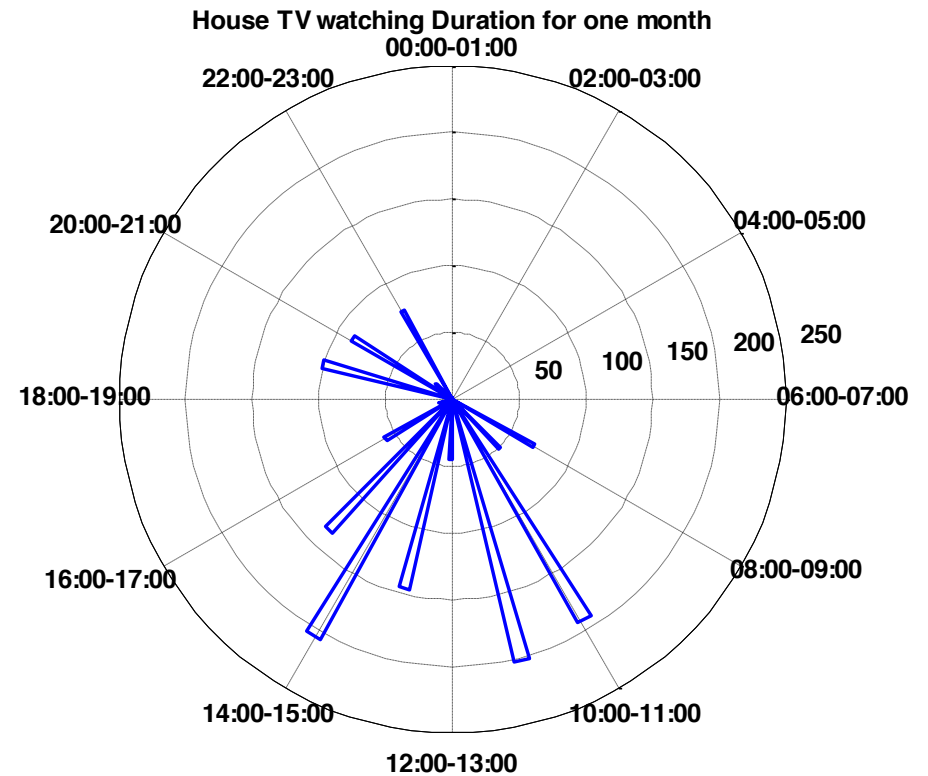
House 4. 2 retired adults

Results 2. Activity time profiles varied between households of similar compositions.



watching TV

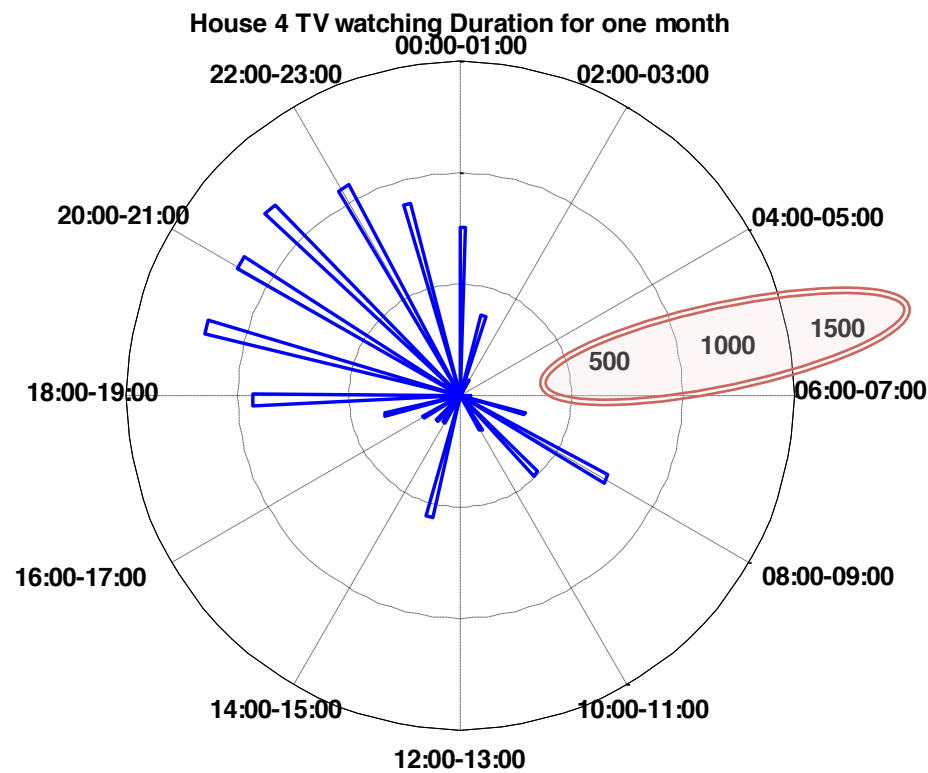
House 4. 2 retired adults



watching TV

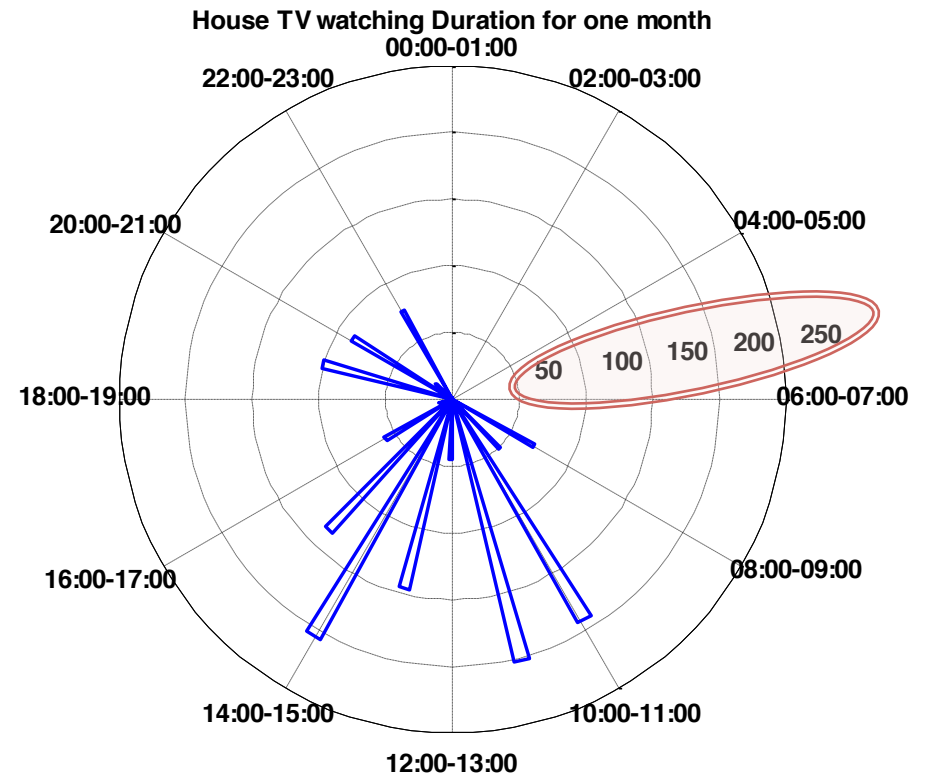
House 8. 2 retired adults

Results 2. Activity time profiles varied between households of similar compositions.



watching TV

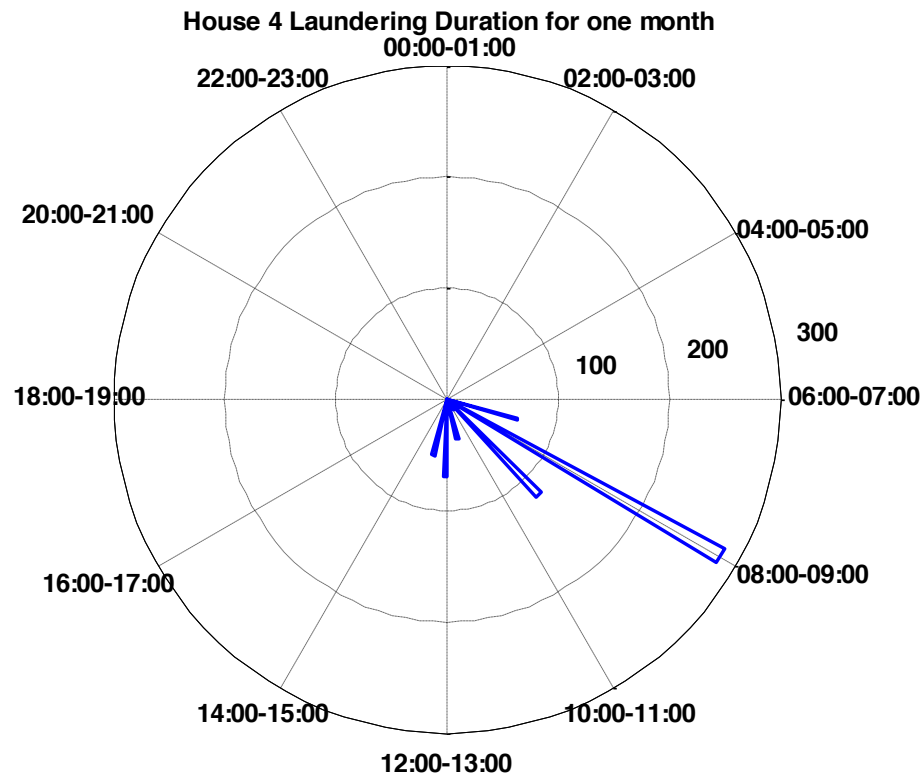
House 4. 2 retired adults



watching TV

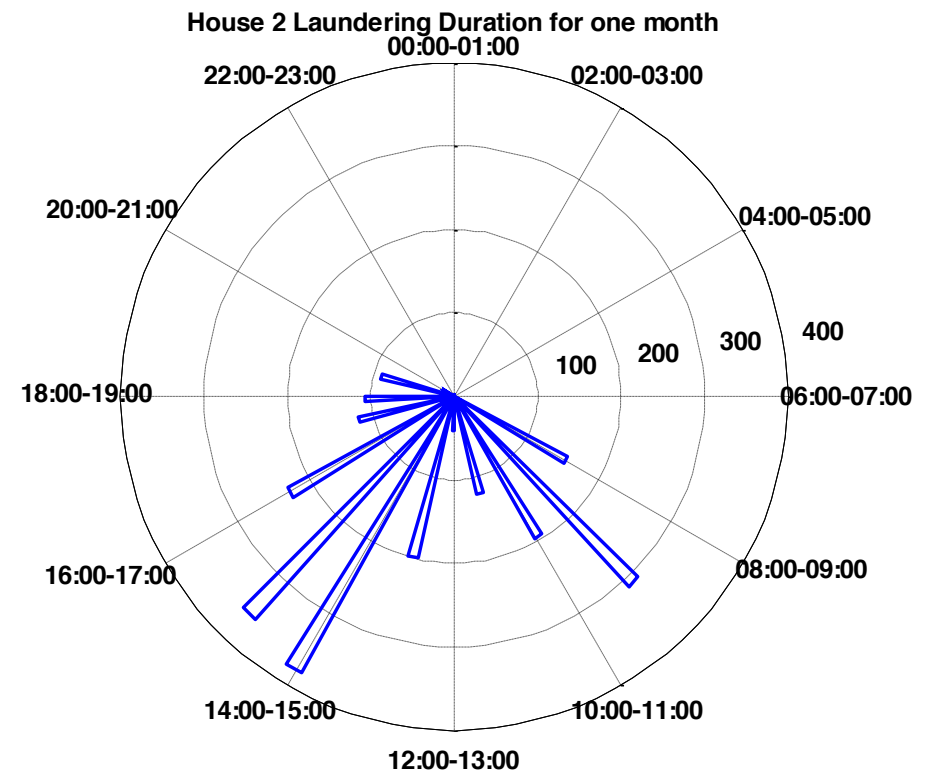
House 8. 2 retired adults

Results 2. Activity time profiles varied between households as a result of **different compositions**.



laundrying

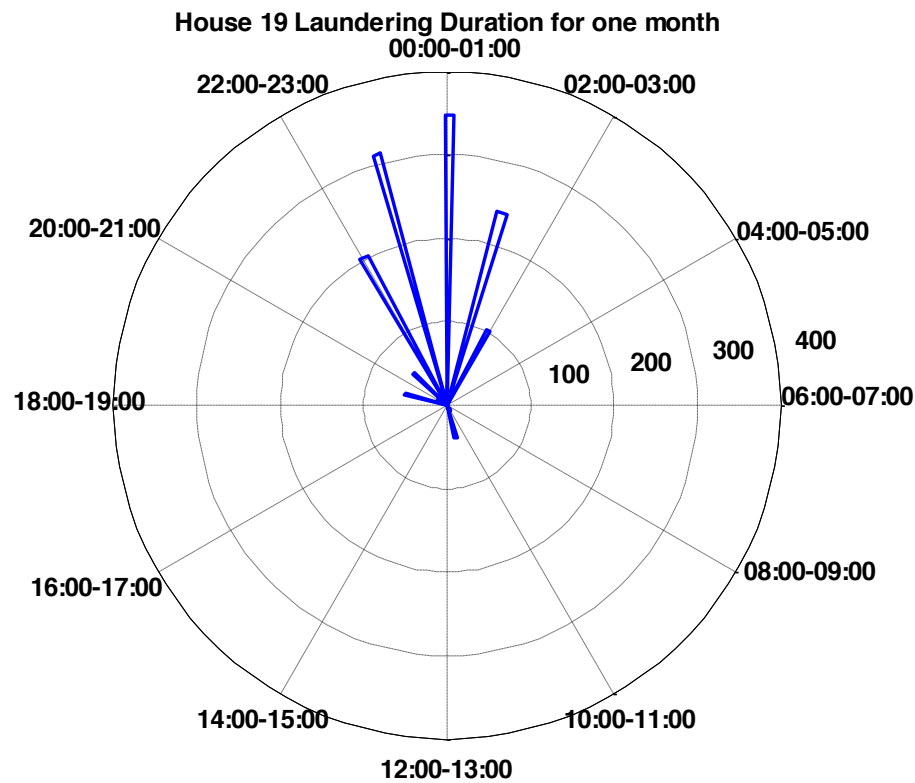
House 4. 2 retired adults



laundrying

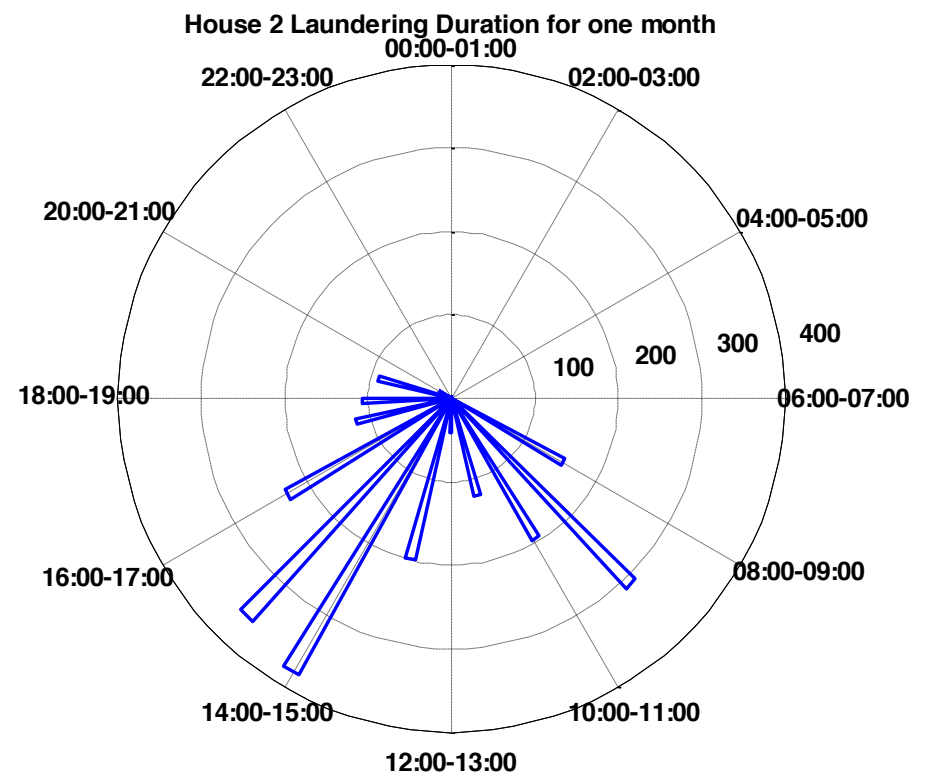
House 2. 2 adults + 2 pre-school children

Results 2. Activity time profiles varied between households with **different scheduling**.



laundrying

House 19. 2 adults + 2 school children



laundrying

House 2. 2 adults + 2 pre-school children

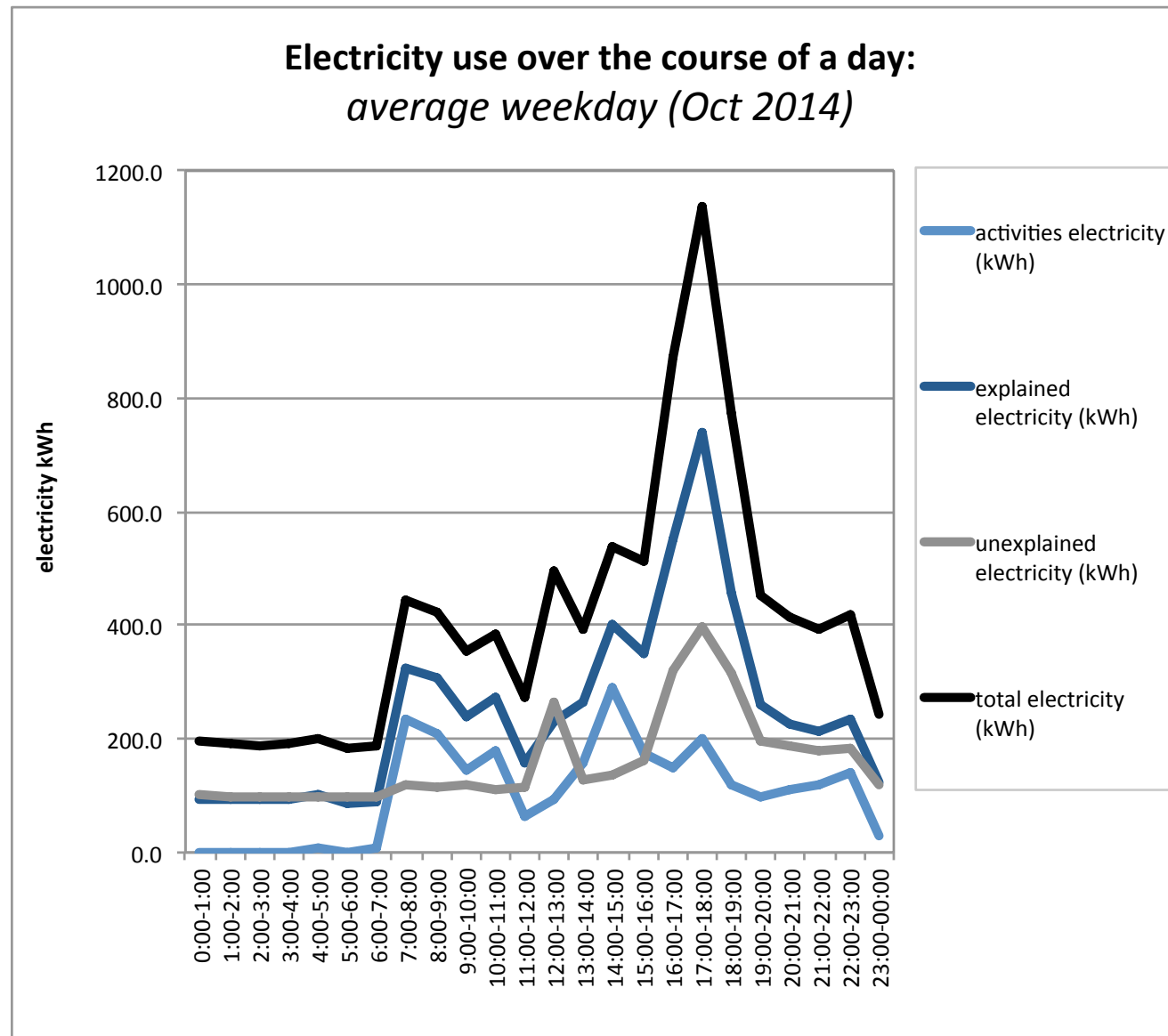
Results 3. Activity time profiles are **inter-linked temporally** to different extents.

correlations (time use): average weekday

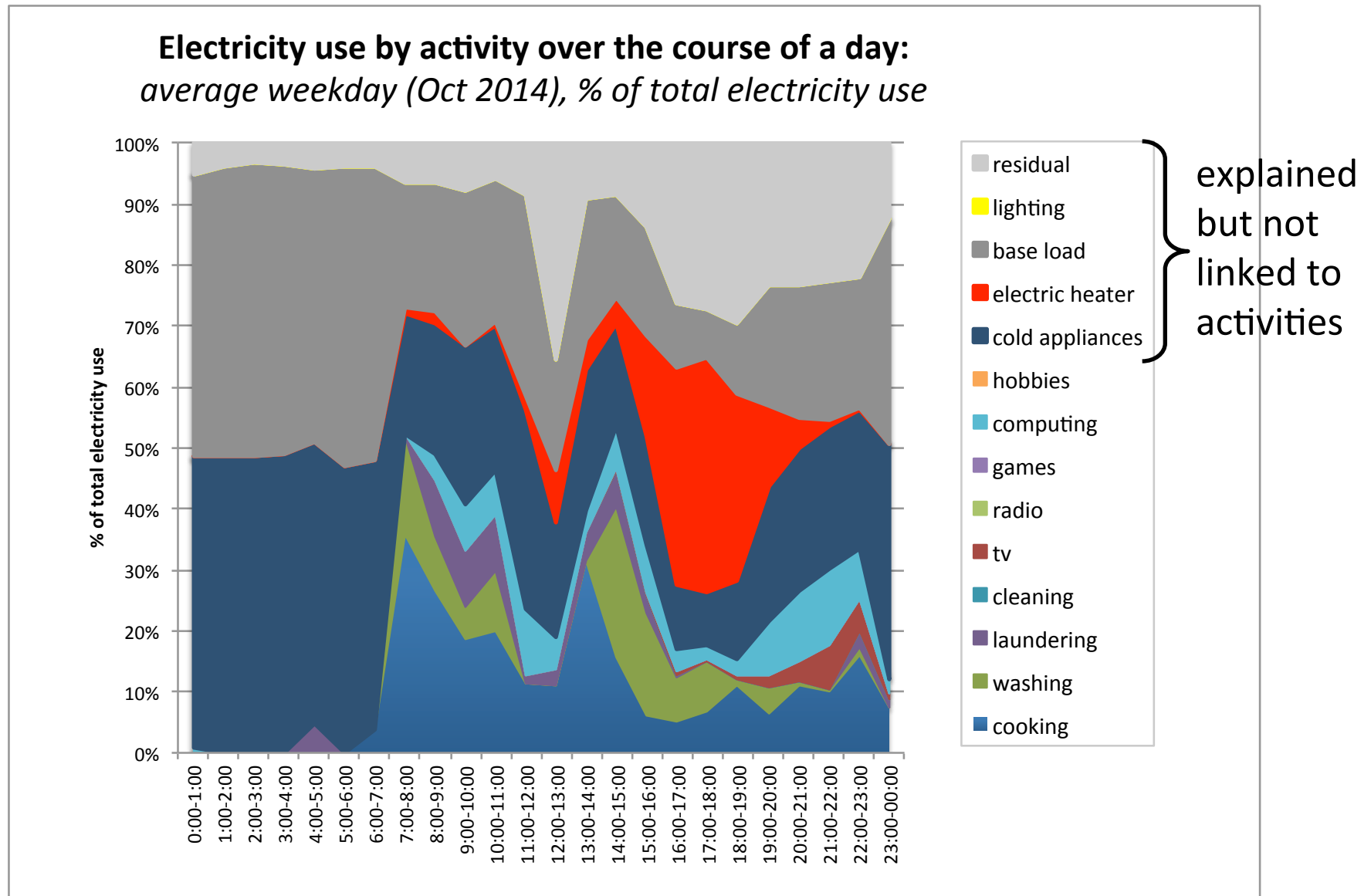
	cooking	eating	laundrying	sleeping	watching tv	computing	hobbies
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eating	0.70	1.00					
laundrying	0.17	0.46	1.00				
sleeping	-0.63	-0.51	-0.22	1.00			
tv	0.27	0.16	-0.03	-0.47	1.00		
computing	0.36	0.38	0.29	-0.56	-0.24	1.00	
hobbies	0.53	0.33	0.28	-0.50	-0.25	0.57	1.00

House 4. 2 retired adults

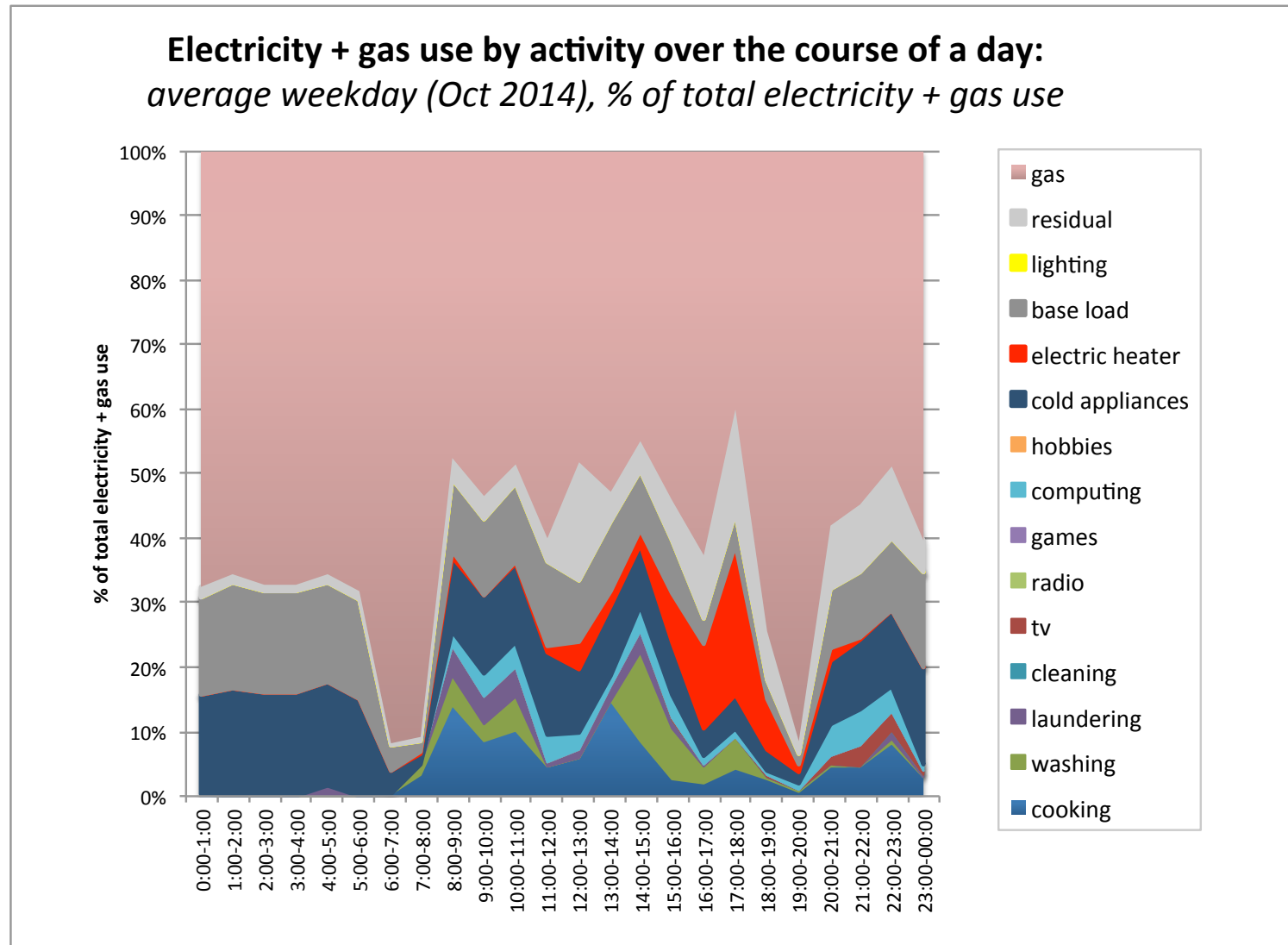
Results 4. Energy use can be apportioned among activities.



Results 4. Energy use can be linked to activities to provide **meaningful feedback** to households.



Results 4. Energy use can be linked to activities to provide **meaningful feedback** to households.



Applications

1. Comparison with time use statistics
-> *identification with national clusters*

Method ... Results ... **Applications**

Applications

1. Comparison with time use statistics
-> *identification with national clusters*
2. Temporal stability and sequencing of activities
-> *demand response*

Method ... Results ... **Applications**

Applications

1. Comparison with time use statistics
-> identification with national clusters
2. Temporal stability and sequencing of activities
-> demand response
3. Activity-based energy feedback
-> retrofit & energy savings

Method ... Results ... **Applications**

Applications

1. Comparison with time use statistics
-> identification with national clusters
2. Temporal stability and sequencing of activities
-> demand response
3. Activity-based energy feedback
-> retrofit & energy savings
4. Reduced methodology for smart meter rollout
-> real-time activity time profiles nationwide!

Method ... Results ... Applications ... **The End**

Applications

1. Comparison with time use statistics
-> identification with national clusters
2. Temporal stability and sequencing of activities
-> demand response
3. Activity-based energy feedback
-> retrofit & energy savings
4. Reduced methodology for smart meter rollout
-> real-time activity time profiles nationwide!

questions? suggestions? + informal session 2-3pm

Identifying Household Activities from Smart Meter Data

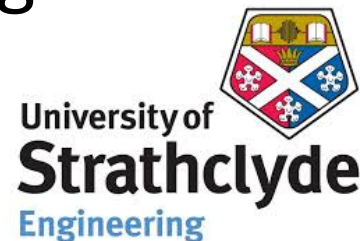
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activities

*things people do,
ways people describe
their day or how they
spend their time*

e.g.

cooking, washing,
resting, caring

descriptive characterisation
of everyday life in a way that
is meaningful to households

practices

*socially-shared bundles
of doings and sayings
constituted by
technologies, skills,
meanings and rules*

e.g.

cooking, washing,
socialising

theory-based analysis of
constitutive elements of
everyday life

energy services

*useful functions
provided by energy
end-use technologies*

e.g.

cooking, washing,
heating, lighting

ultimate purposes for
which energy is consumed
as part of everyday life

Method: Step 1. Develop a comprehensive set of activities to capture household life *and* energy use.

DAILY ROUTINES

cooking cooking, preparing food & drink, washing up

eating eating, drinking

washing showering, washing, dressing

laundry doing laundry

cleaning cleaning, housework, (other than laundry or washing up)

sleeping sleeping, resting

INTERACTING

communicating communicating, interacting with people outside the home

socialising entertaining, socialising, being with people at home

LEISURE & ICTs

tv watching tv, video, film or any audiovisual

radio listening to radio, music or any audio

games playing games on console, computer, tablet, smartphone

computing using computer, tablet, smartphone, (other than for games or work)

OTHER ACTIVITIES

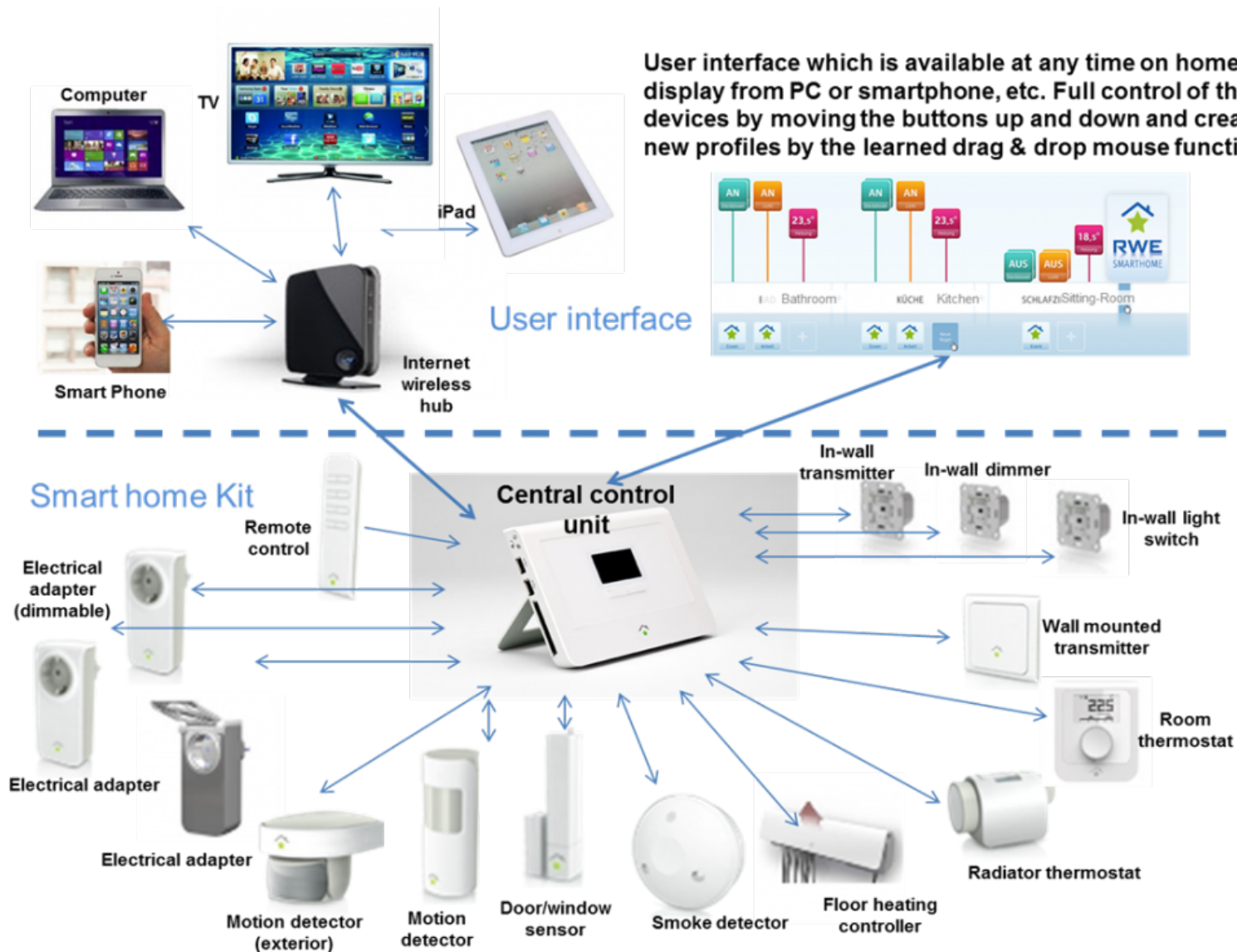
hobbies doing hobbies, sports

caring looking after children, caring for household members

working working, studying

other other activities

A smart home collects data from meters and sensors, and provides households with new control functions



quant.

qual.

mixed

Data requirements

time use studies

household interviews and
video ethnography on
activities and technologies

household interviews and
video ethnography

house survey
(appliances, rooms)

electricity & gas smart meters

appliance / plug monitor

environmental sensors

self-completion
appliance time diaries

self-completion
activity time diaries

Methodology, step-by-step

1. Define activities

*comprehensive set of activities
meaningful to households*

2. Build activity ontology

*map relationships between
technologies and activities*

3. Collect real-time data

*collect & process energy and
environmental data from homes*

4. Disaggregate data

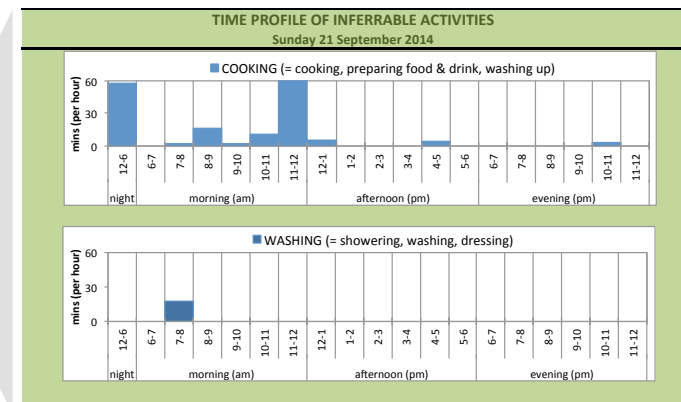
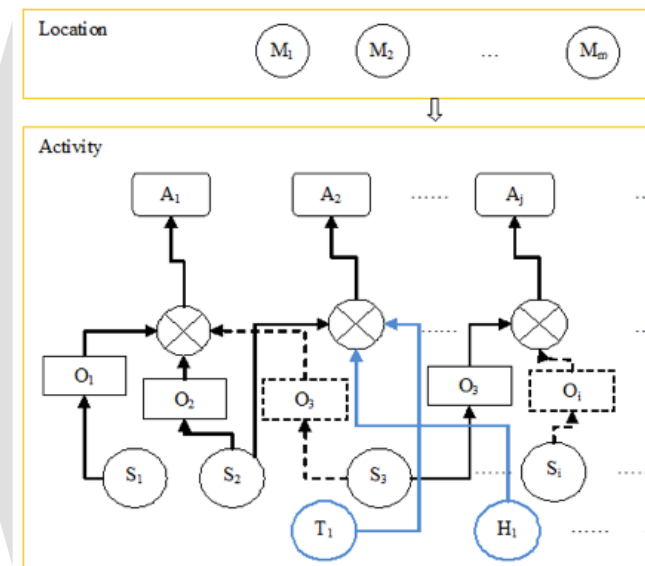
5. Infer activities

*process data using activity
ontology and inference algorithms*

6. Validate inferences

*compare inferences against time
diary data & validate with household*

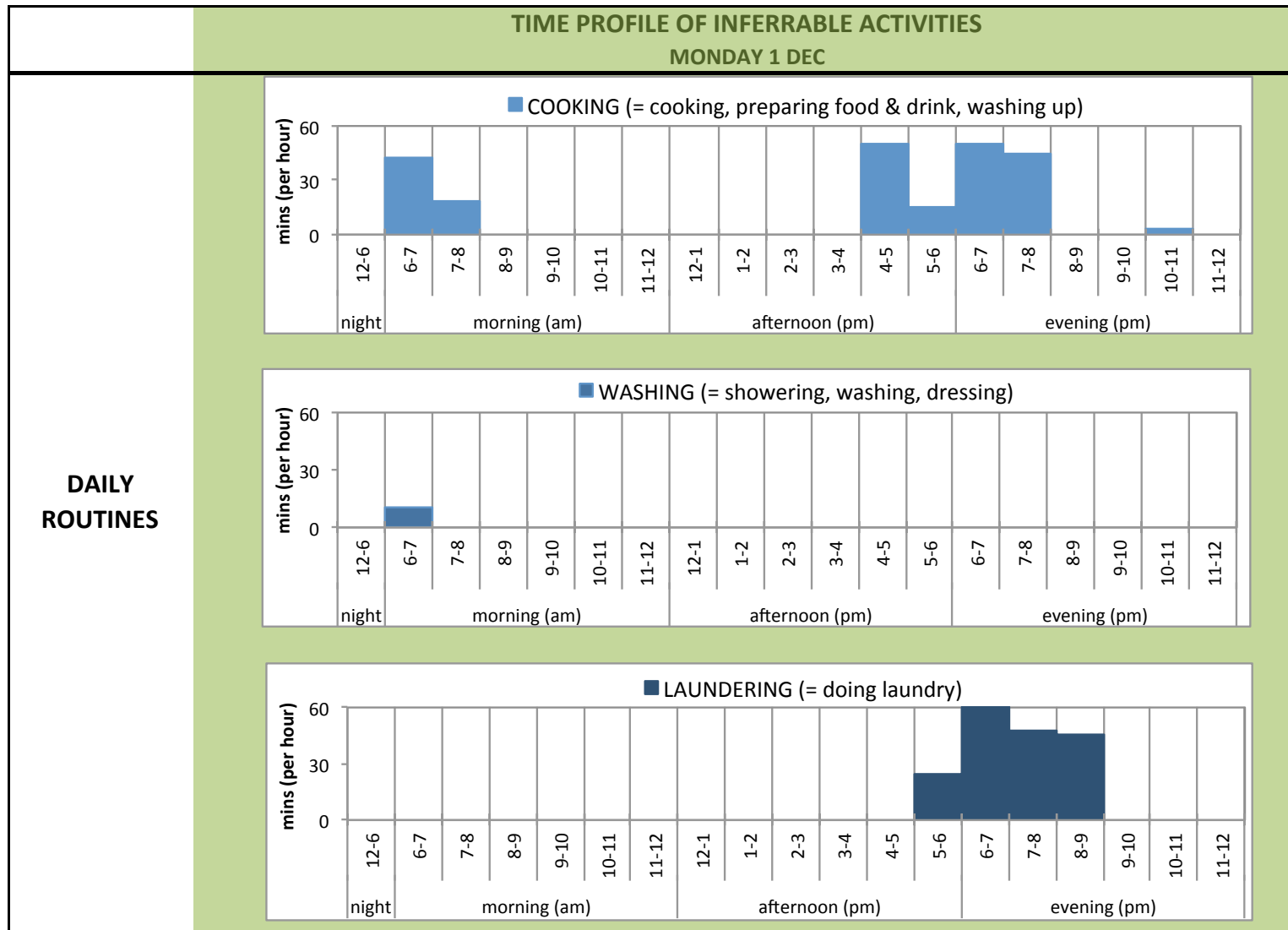
Illustrative examples



1-6-2014
DATE: NUMBER OF PEOPLE AT HOME: 2

	MORNING ACTIVITIES (until noon)	early -> 6.00 ->	7.00 -> 8.00 ->	9.00 -> 10.00 ->
cooking, preparing food & drink, washing up	empty dishwasher	7.10 - 7.25	8.25 - 8.40	
eating, drinking		7.40 - 7.55		9.00 - 9.05
showering, washing, dressing		7.05 - 7.30		
sleeping, resting		7.20 - 7.30		9.00 - 9.20
working, studying				

Method: *Step 5*. Make **inferences** about time profile of activities (process energy data through ontology).



Method: *Step 6*. Validate activity inferences using activity **time diaries**.

1-6-2014

DATE: NUMBER OF PEOPLE AT HOME: 2.

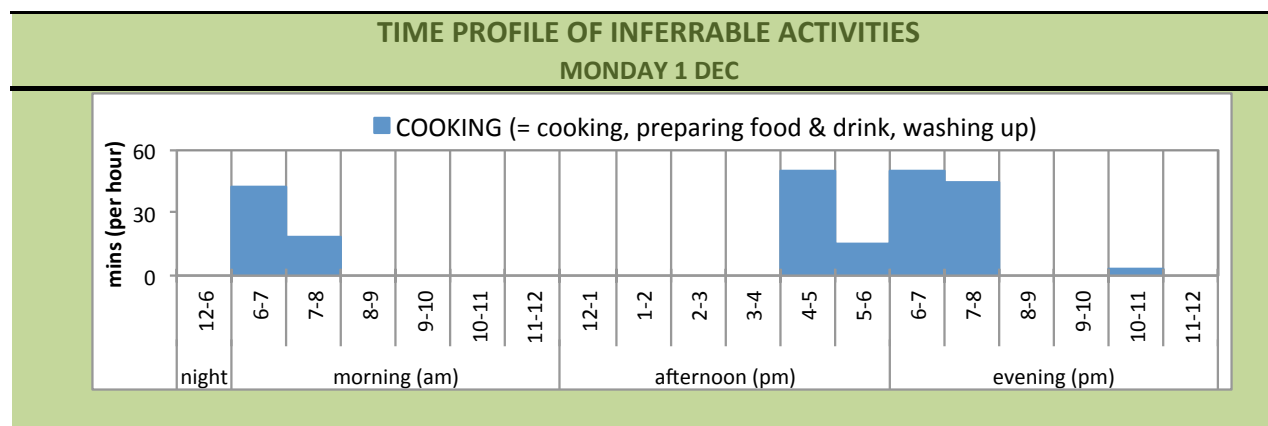
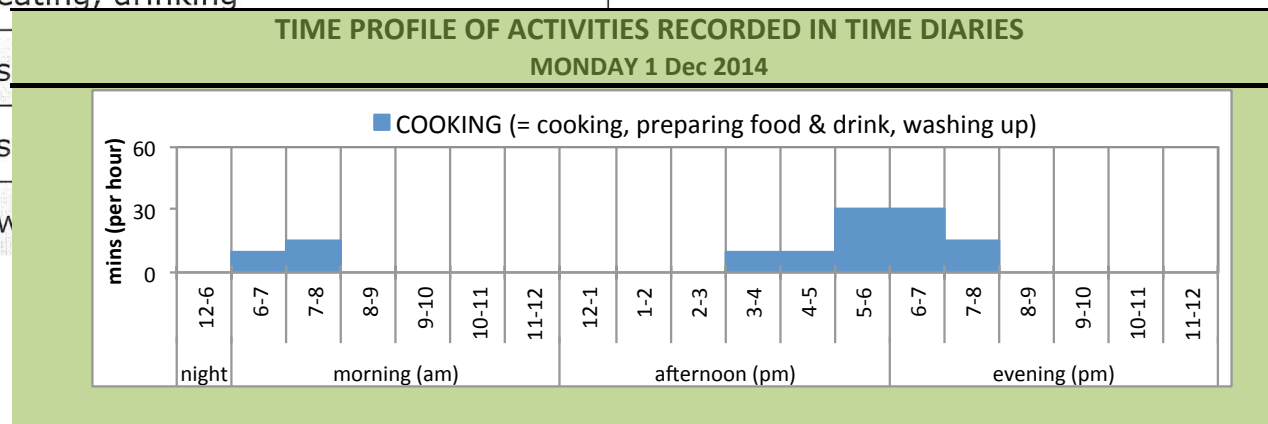
	MORNING ACTIVITIES (until noon)	early ->	6.00 ->	7.00 ->	8.00 ->	9.00 ->	10.00
DAILY ROUTINES	cooking, preparing food & drink, washing up <i>empty dishwasher</i>			7-10	- - - 8.25		
	eating, drinking			7.40-7.55		9.00-9.05	
	showering, washing, dressing			R 7.05-7.20 A 7.20-7.30			
	sleeping, resting			sleep until 6.00.			9.00-9.20
	working, studying						

Method: *Step 6. Validate* activity inferences using activity time diaries.

1-6-2014

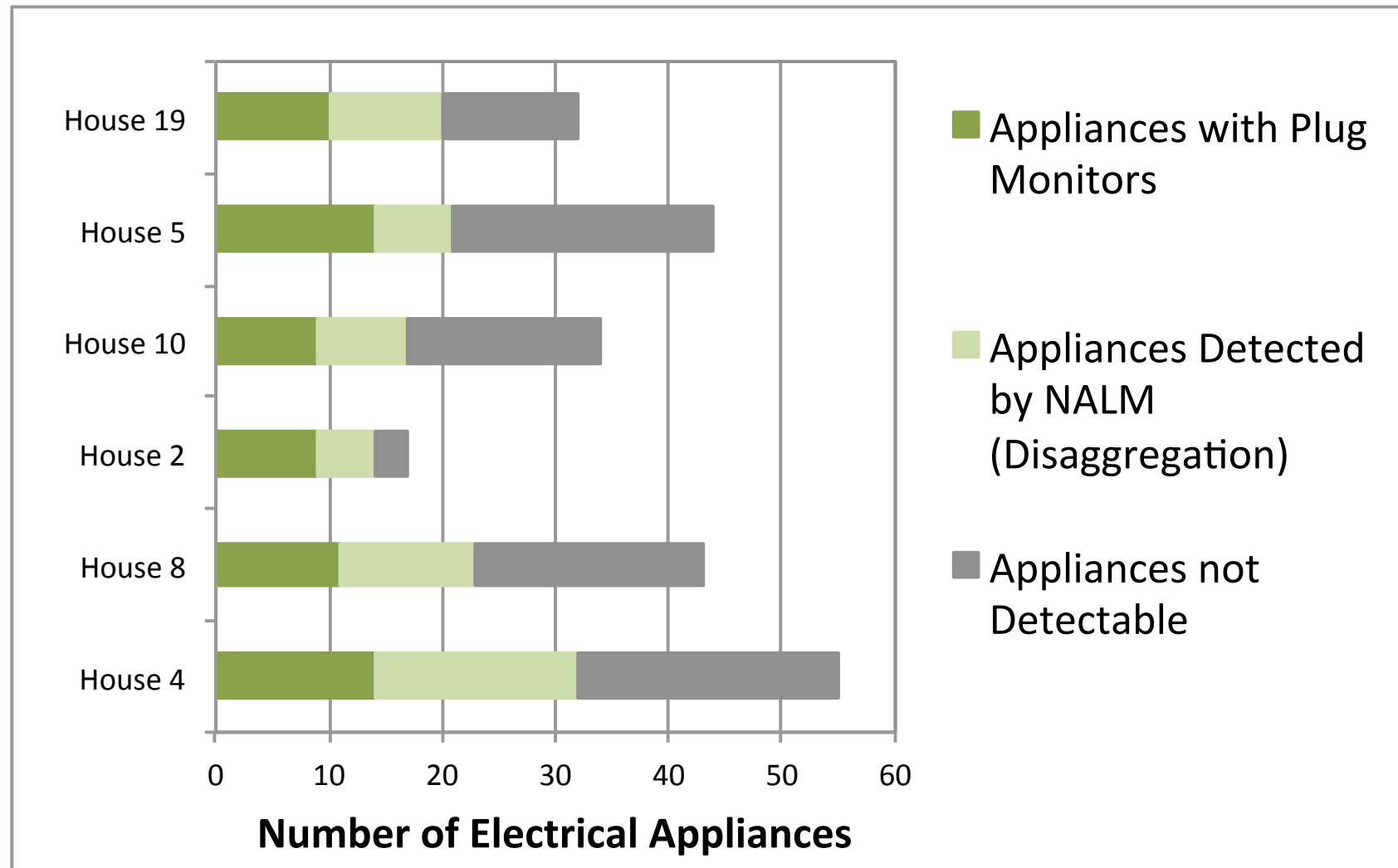
DATE: NUMBER OF PEOPLE AT HOME: 2.

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	eating, drinking			7.40-7.55		9.00-9.05	



adjust
ontology
or
appliance
detection

Results 1. 50-80% of electrical appliances were **detectable** (exceptions: low power, battery/mobile).



Results 3. Activity time profiles are **inter-linked temporally** to different extents.

correlations (time use): average weekday

	cooking	eating	laundrying	sleeping	watching tv	computing	hobbies
cooking	1.00						
eating	0.70	1.00					
laundrying	0.17	0.46	1.00				
sleeping	-0.63	-0.51	-0.22	1.00			
tv	0.27	0.16	-0.03	-0.47	1.00		
computing	0.36	0.38	0.29	-0.56	-0.24	1.00	
hobbies	0.53	0.33	0.28	-0.50	-0.25	0.57	1.00

correlations (time use): average weekend day

	cooking	eating	laundrying	sleeping	watching tv	computing	hobbies
cooking	1.00						
eating	0.75	1.00					
laundrying	0.07	0.43	1.00				
sleeping	-0.54	-0.60	-0.19	1.00			
tv	0.17	0.33	0.26	-0.42	1.00		
computing	-0.18	-0.12	0.01	-0.23	0.06	1.00	
hobbies	0.03	0.38	0.92	-0.20	0.19	0.24	1.00

House 4. 2 retired adults

Results 3. Activity time profiles are **inter-linked temporally** to different extents.

correlations (time use): average weekday

	cooking	eating	laundrying	sleeping	watching tv	computing	hobbies
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correlations (time use): average weekend day

	cooking	eating	laundrying	sleeping	watching tv	computing	hobbies
cooking	1.00						
eating	0.75	1.00					
laundrying	0.07	0.43	1.00				
sleeping	-0.54	-0.60	-0.19	1.00			
tv	0.17	0.33	0.26	-0.42	1.00		
computing	-0.18	-0.12	0.01	-0.23	0.06	1.00	
hobbies	0.03	0.38	0.92	-0.20	0.19	0.24	1.00

House 4. 2 retired adults