

Fraunhofer Center for Sustainable Energy Systems

Non-Intrusive Appliance  
Load Monitoring (NIALM):  
Review and Outlook\*

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Electronics*

*Las Vegas*

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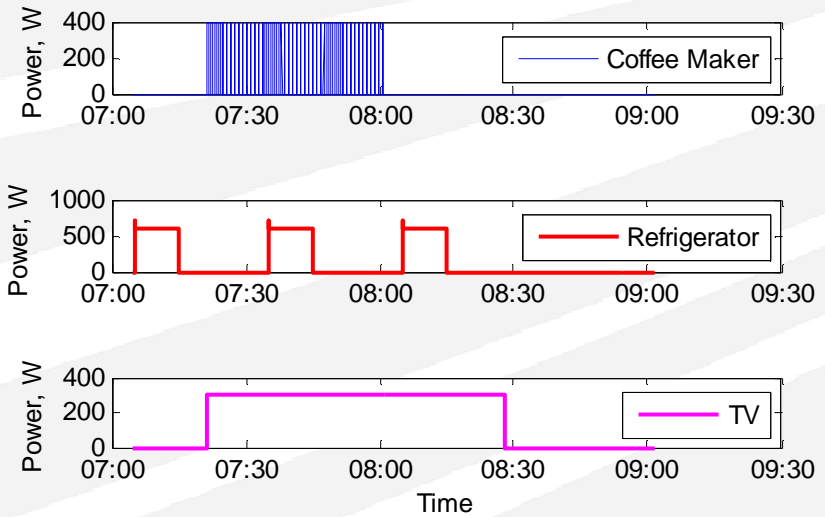
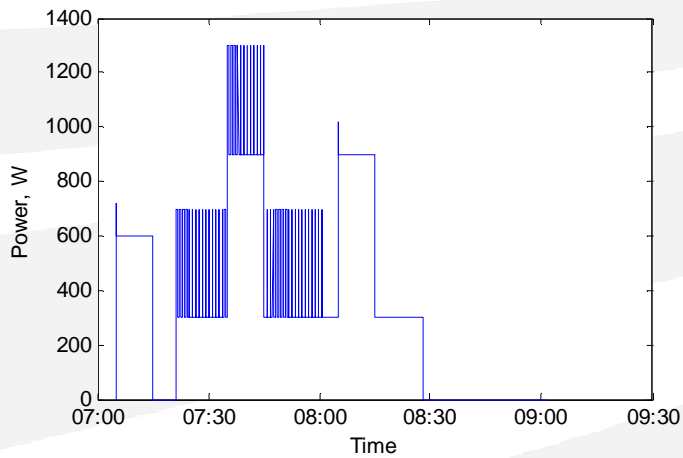
# Fraunhofer: A Leading Force in Applied R&D

- **Fraunhofer Gesellschaft:** Europe's largest contract research organization – 17,000 employees
- Annual research budget: US\$2.2B
- 59 institutes provide contract R&D and certification services for clients
- **Research Themes:** health, security, communication, mobility, energy, environment
  
- **Fraunhofer USA:** 7 Centers, >200 employees, Annual research budget >\$50M
- Fraunhofer CSE is located in Cambridge, MA
- Three focus areas:
  - Building Energy Efficiency
  - PV Modules
  - Smart Grid



# What Is NIALM?

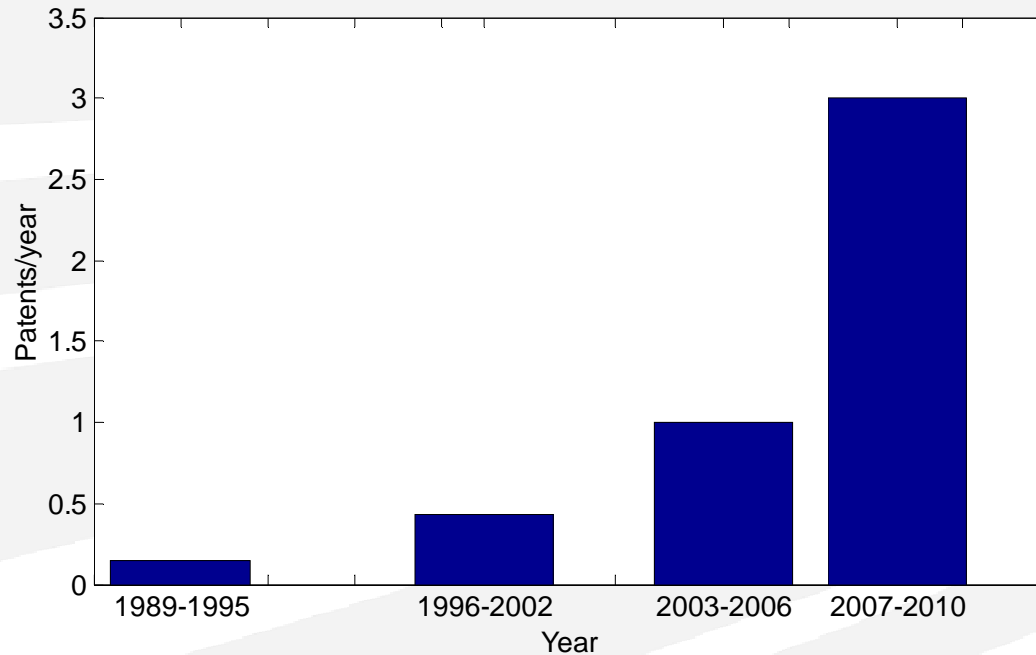
- Non-Intrusive Appliance Load Monitoring
  - A.k.a. Non-Intrusive Load Monitoring



- Main breaker/circuit level
- Data acquisition (hardware) and disaggregation algorithms (software)

## NIALM: Interest

- Number of US granted patents – exponential growth

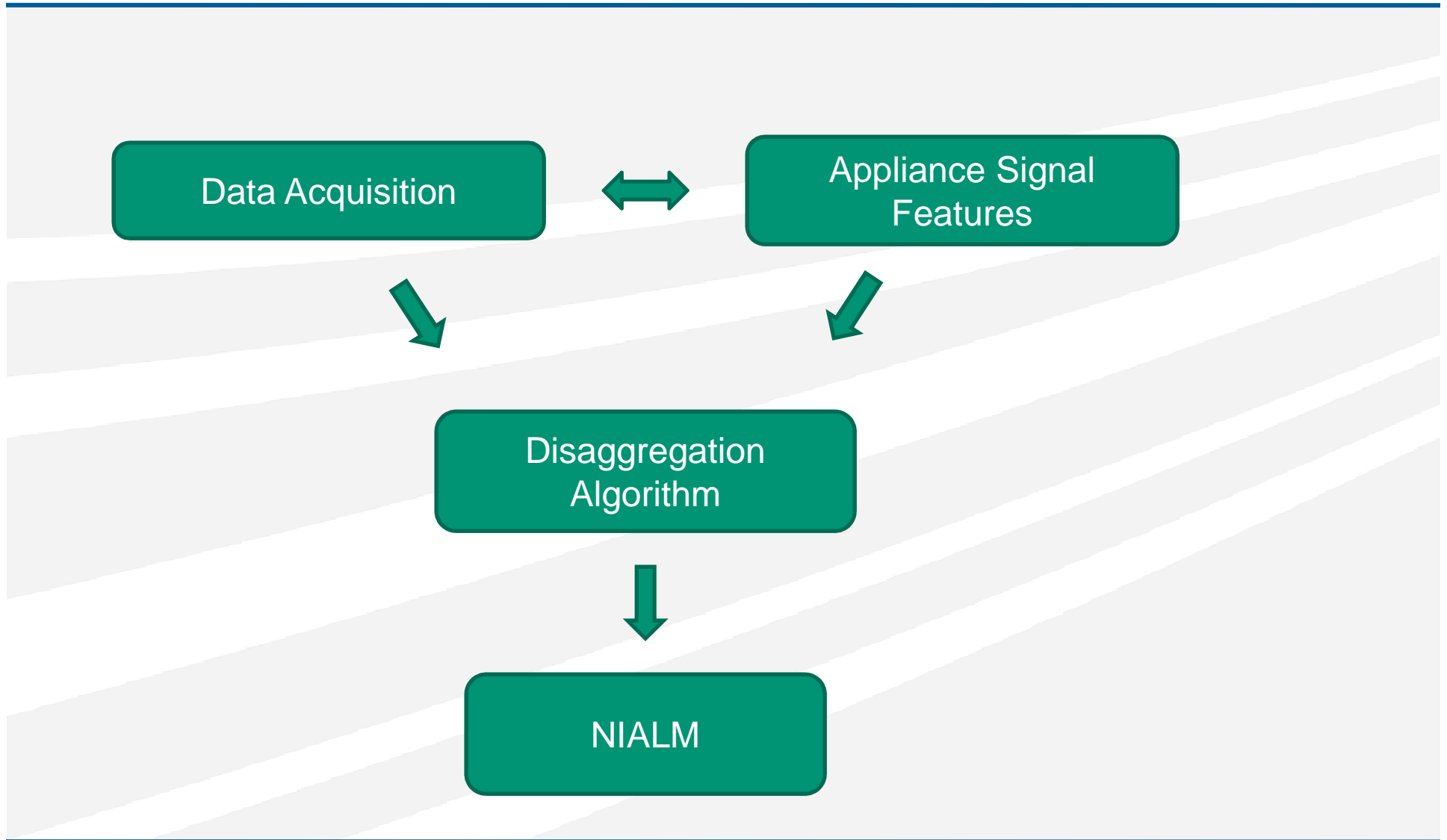


- Both large (Belkin, GE, IBM, Intel,) and small (4home, Emme, Enetics, Navetas) companies

## NIALM: Motivation

- Energy saving potential
  - Residential buildings consume ~ 20% of primary energy in USA
  - Electricity constitutes ~ 70% of residential energy consumption
  - Feedback on energy usage can result in 5-15% saving
- Smart Grid – unique business opportunity
  - ~ 80 M smart meters in USA are projected by 2013
  - Hardware can be used by NIALM
  - Non-intrusive “manual” alternative to home automation
- Time-of-use electricity pricing
  - Transition from fixed to time-dependent pricing in Europe and North America
  - More people are concerned with energy monitoring and management

# How Does NIALM Work?



# Data for NIALM: Sampling Rate

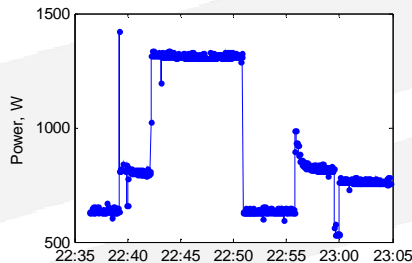
Typical daily number of switching events in a household: up to  $\sim 10^3$ - $10^4$  [Baranski, 2004]



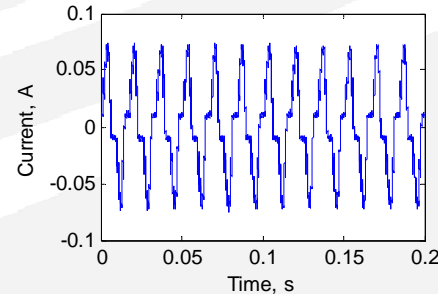
0.1 Hz – minimum sampling rate



Inexpensive hardware:  
1 Hz



More expensive hardware:  $> 5$  kHz



# NIALM: Types of Appliances

Permanent



On/off



Multi State



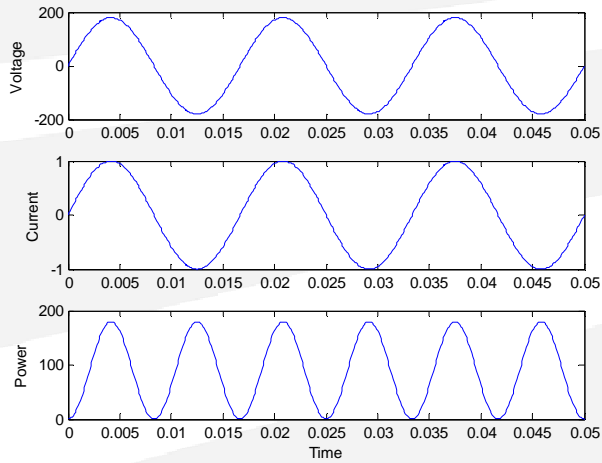
Variable



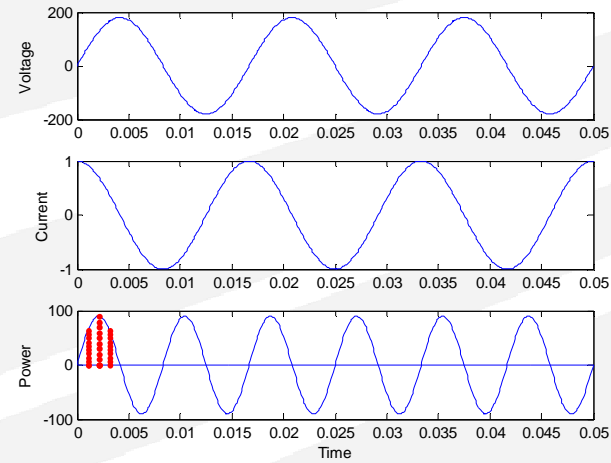


# I. Low Sampling Rate, Features: Power Change

## Real Power

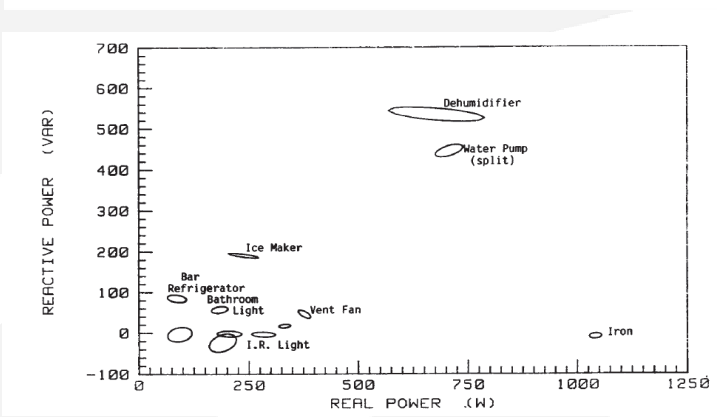
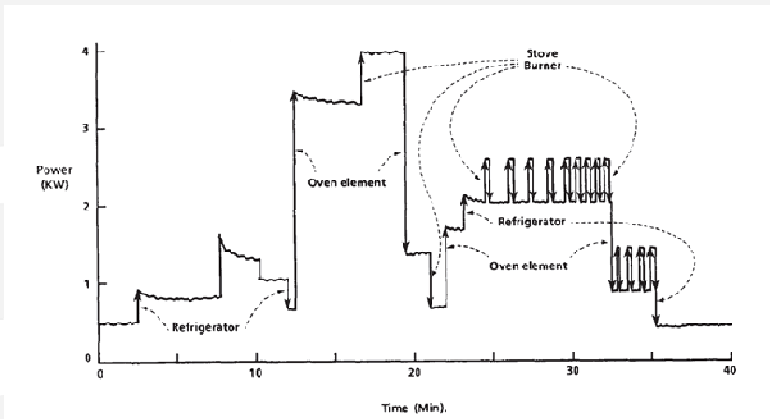


## Reactive Power



# I. Power Change: Basic Method

Hart, 1992 (MIT method)



Accuracy: ~80%

On/off appliances

Multi State

Variable Load

Permanent Load

Same-Load Appliances

# I. Change of Real Power Only

Baranski and Voss, 2003-2004

□ Hart's Method:  
single  $-\Delta P \leftrightarrow$  single  $+\Delta P$

□ Baranski's Method:  
multiple  $-\Delta P_i \leftrightarrow$  multiple  $+\Delta P_i$

□ Baranski's algorithm – error term minimization  
(optimization by genetic algorithm)

On/off  
appliances

Multi State

Variable  
Load

Permanent  
Load

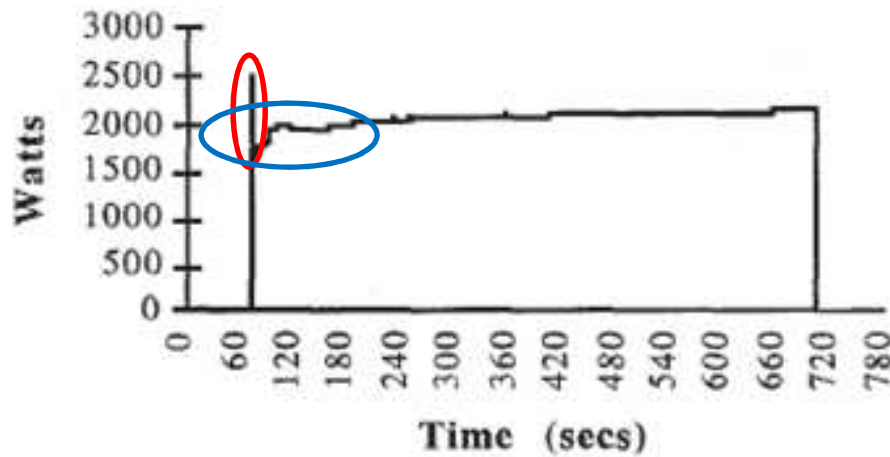
Same-Load  
Appliances

Accuracy: ~85%

# I. Change of Power + Additional Features

Albicki and Cole, 1998

Edges and slopes as additional features



On/off  
appliances

Multi State

Variable  
Load

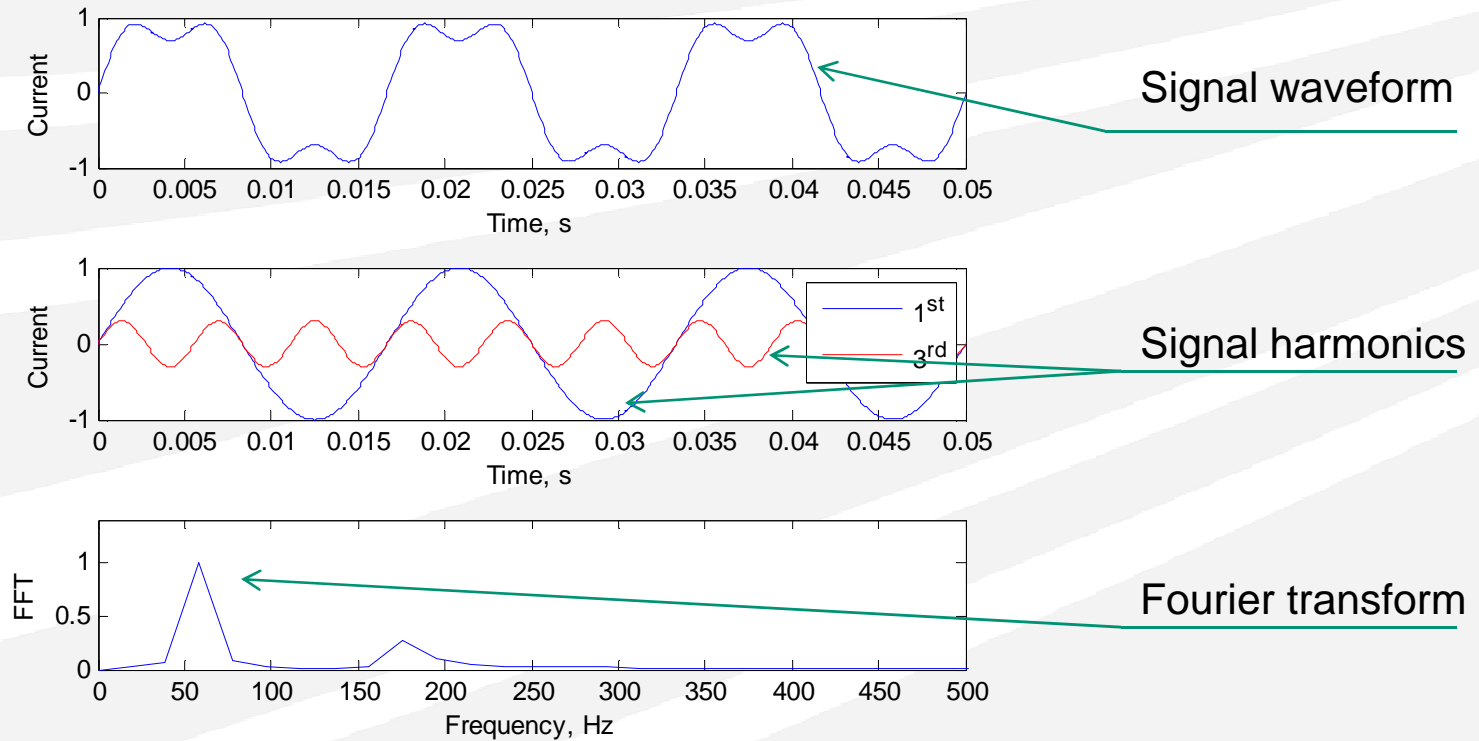
Permanent  
Load

Same-Load  
Appliances

Accuracy: ~N/A

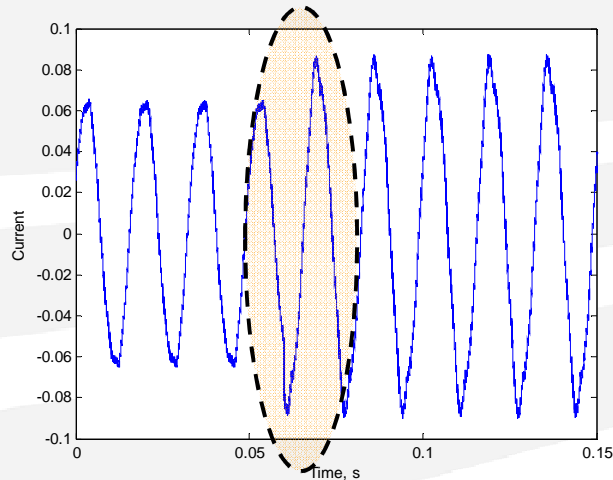
## II. High Sampling Rate, Harmonics as Features

### Basic Idea: Fourier Transform



## II. Features: Harmonics for Transients (1/2)

Laughman *et al.*, 2003 (MIT)



- 1) Detect transients
- 2) Subtract pre-transient from post-transient
- 3) Take Fourier transform and compare with template
- 4) Set of harmonics = “power envelope”

Accuracy: ~N/A

On/off  
appliances

Multi State

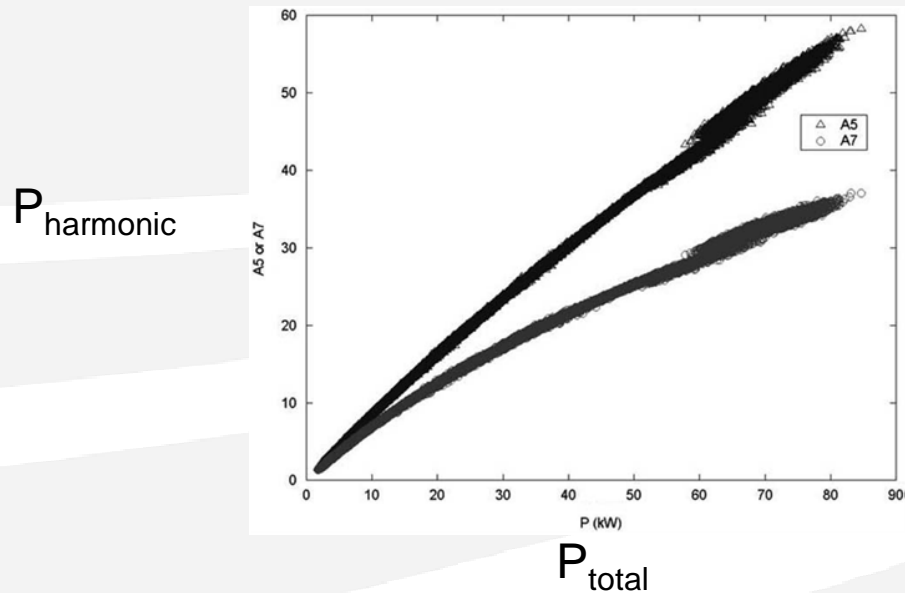
Variable  
Load

Permanent  
Load

Same-Load  
Appliances

## II. Features: Harmonics for Transients (2/2)

Lee *et al.*, 2005 (MIT)



- 1) Strong correlation between harmonics and power draw
- 2) Estimate harmonics then estimate load

Accuracy: ~N/A

On/off  
appliances

Multi State

Variable  
Load

Permanent  
Load

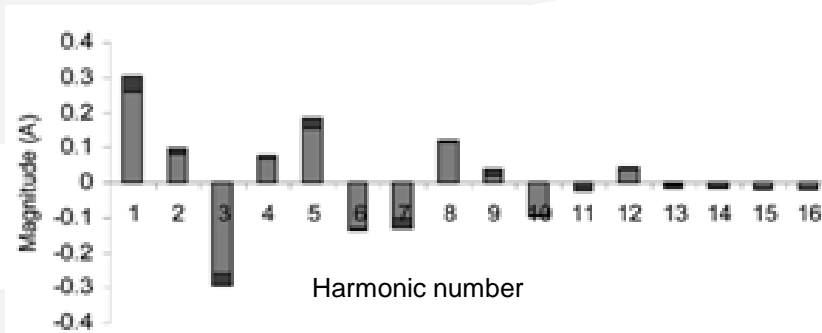
Same-Load  
Appliances

## II. Features: Harmonics for Steady and Transients

Srinivasan *et al.*, 2006 (Singapore)

Basic idea: Set of harmonics for  
*all combinations* of appliances

- 1) Obtain Fourier transform continuously
- 2) Compare with templates



Accuracy: 77% (phone charger) – 99% (PC)

On/off  
appliances

Multi State

Variable  
Load

Permanent  
Load

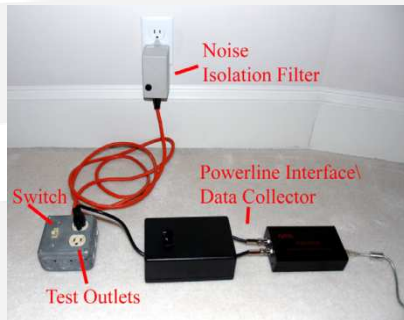
Same-Load  
Appliances



## II. Features: Fourier Transform of Noise (1/2)

Patel *et al.*, 2007 (Ga. Tech.)

Basic idea: appliance connected to a socket induces noise (electromagnetic interference) in another socket



- 1) Measure signal at high rate (500 kHz)
- 2) Calculate Fourier transform continuously
- 3) Compare with templates

Accuracy: 85-90%

On/off  
appliances

Multi State

Variable  
Load

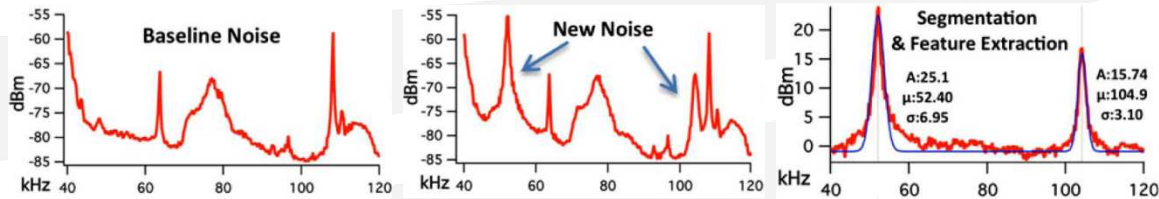
Permanent  
Load

Same-Load  
Appliances

## II. Features: Fourier Transform of Noise (2/2)

Patel *et al.*, 2010 (Now U. Wash.)

Basic idea: switch mode power supplies (SMPS) generate distinctive noises



- Distinctive features (“applicable across homes”)
- Yet features variable enough to “differentiate between similar devices in a home”

On/off  
appliances

Multi State

Variable  
Load

Permanent  
Load

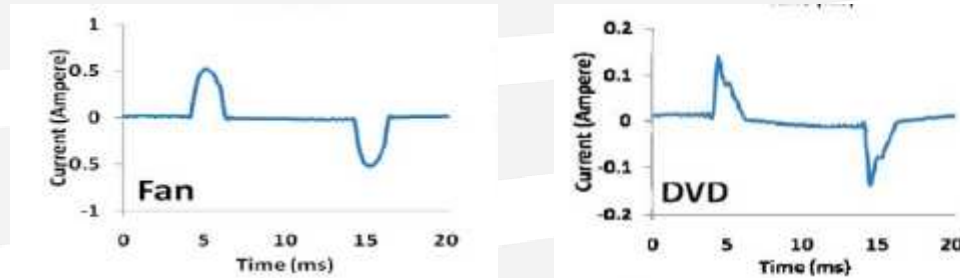
Same-Load  
Appliances

Accuracy: up to 93.8%

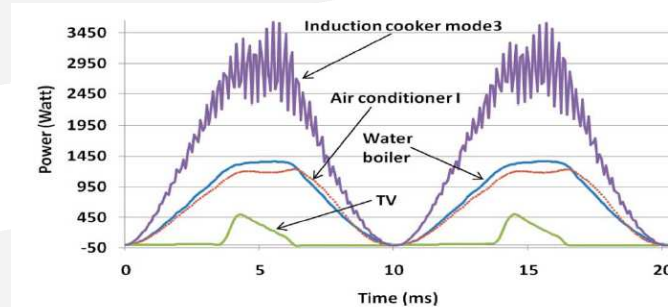
## II. Features: Beyond Harmonics / Fourier Transform

Liang *et al.*, 2010 (China Light & Power)

Waveform:



Power waveform



Accuracy: 92.7%

On/off  
appliances

Multi State

Variable  
Load

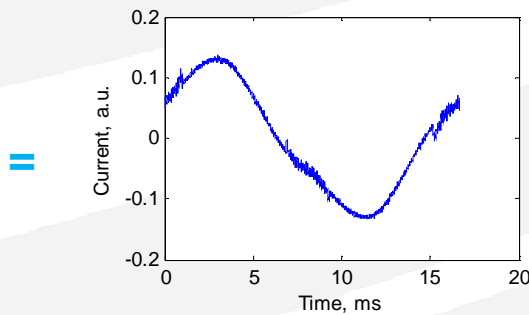
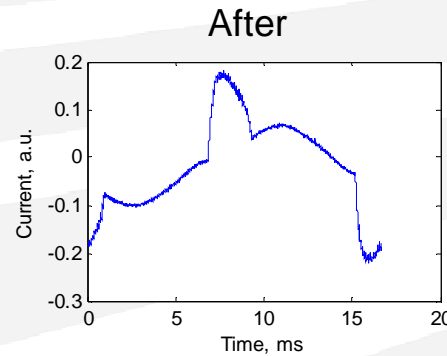
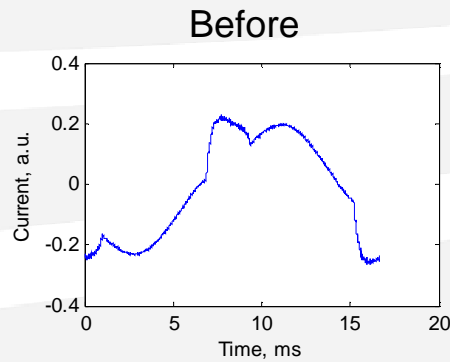
Permanent  
Load

Same-Load  
Appliances

# NIALM Algorithms (1/3)

## Single-matching (pattern recognition)

Example: Waveform



Pattern recognition methods used:

- N-nearest neighbor
- Bayes classifier
- Neural networks

On/off appliances

Multi State

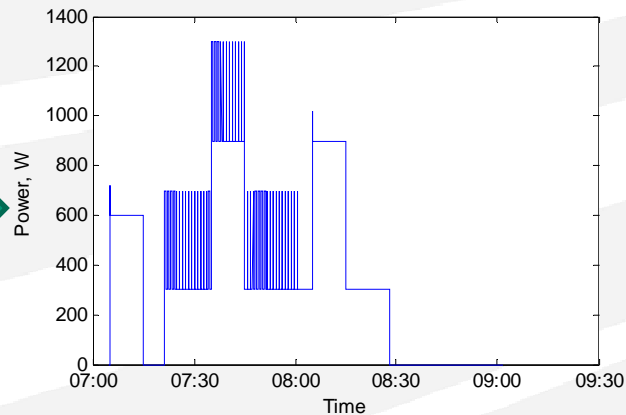
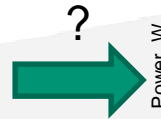
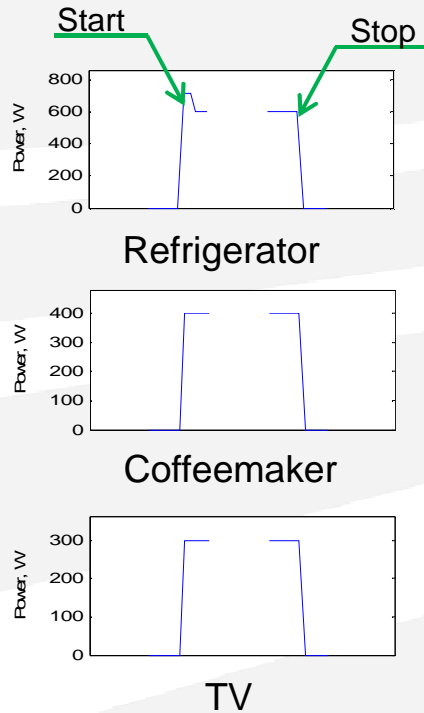
Variable Load

Permanent Load

Same-Load Appliances

# NIALM Algorithms (2/3)

## Multiple-matching (Optimization)



On/off appliances

Multi State

Variable Load

Permanent Load

Same-Load Appliances

Optimization methods used:

- Least squares
- Integer programming
- Genetic algorithms

# NIALM Algorithms (3/3)

## True disaggregation

A.k.a. blind source separation

- Possible methods:
  - Independent Component Analysis (ICA)
  - Singular Value Decomposition
  - Non-negative Matrix Factorization
- Yet has to be tried!

On/off  
appliances

Multi State

Variable  
Load

Permanent  
Load

Same-Load  
Appliances

# Promising Directions

➤ Combination of various “orthogonal” features

On/off  
appliances

➤ Combination of “orthogonal” algorithms

Multi State

• Only one publication so far (Liang *et al.*, 2010)

❖ **The curse of false positives**

Variable  
Load

➤ Use of techniques known elsewhere

Permanent  
Load

• Information Fusion – Dempster-Shafer  
Evidence Theory

Same-Load  
Appliances

## NIALM Accuracy Metrics

- ❑ Only one paper discusses: Liang *et al.*, 2010
- ❑ Can be appliance-related or overall

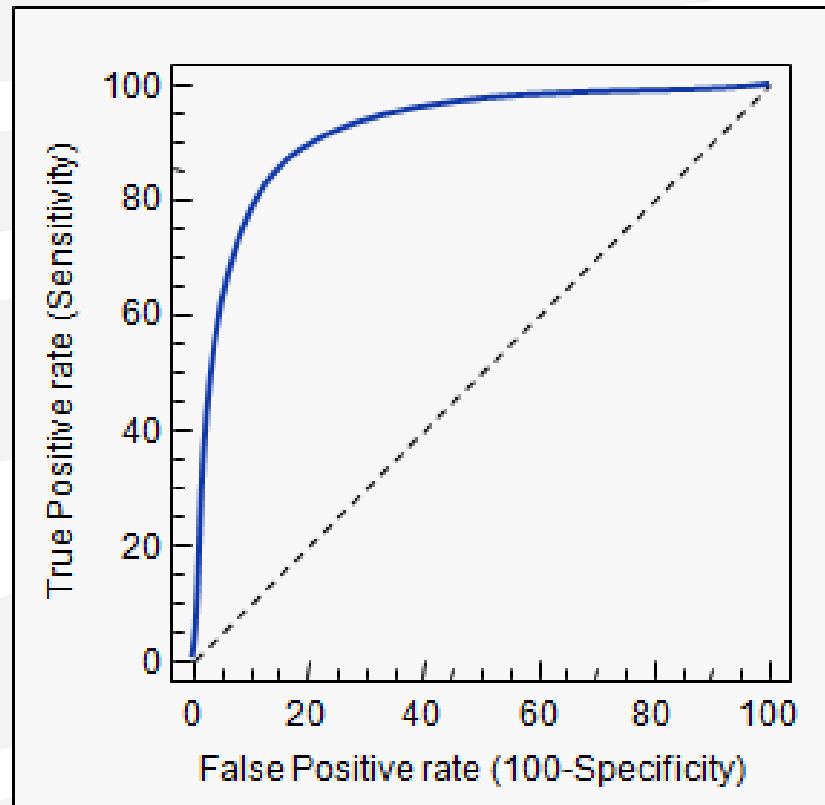
### *Example:*

- Devices were turned on 80 times over monitoring period.
  - 70 out of 80 – detected as “events” (no classification yet) AND also 15 false positive detections.
  - Out of 70, 50 were correctly classified.
  - What is the accuracy???
- Detection accuracy:  $50/(70+15) = 58.8\%$
  - Classification accuracy:  $50/70 = 71.4\%$
  - Overall accuracy:  $50/80 = 62.5\%$



# NIALM Accuracy Metrics – ROC

We suggest: Receiver operating characteristic (ROC)



## Conclusions and Future Trends

- ❑ No complete NIALM solution is available
- ❑ No complete set of robust, widely accepted appliance features is available
- ❑ Using more “orthogonal” features improves accuracy, albeit with higher false positive rates
- ❑ Using several “orthogonal” disaggregation algorithms may improve accuracy, but optimal fusion needs to be implemented
- ❑ ROC curves could be used for algorithm benchmarking rather than the ambiguous “accuracy”

## Our (Fraunhofer CSE) Research Directions

- ❑ Exploration of device “signatures” to identify new orthogonal features
- ❑ Development of identification/tracking algorithms suitable for the new features
- ❑ Optimal fusion of new algorithm(s) with the algorithms available for traditional features
- ❑ Testing and benchmarking – device, whole home & branch circuit levels
- ❑ Exploration of transient event detection techniques
- ❑ Exploration of signal processing techniques for reconstruction of sources
- ❑ Results to be presented at:
  - “Viterbi Algorithm with Sparse Transitions (VAST) for Nonintrusive Load Monitoring,” IEEE Symposium Series on Computational Intelligence, Paris, 2011
  - “Nonintrusive appliance load monitoring (NIALM) for energy control in residential buildings,” Energy Efficiency in Domestic Appliances and Lighting Conference, Copenhagen, 2011