

PresenceSense: Zero-Training Algorithm for Individual Presence Detection based on Power Monitoring

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Experiments

Ultrasonic: measures distance to nearest obstacle **•** For elapse time $\Delta t, d = \frac{1}{2}v_{\text{sound}}\Delta t$ **Acceleration:** detects chair motion Feature $std(\bar{a}(t))$ where $\bar{a}(t) = \sqrt{a_1(t)^2 + a_2(t)^2 + a_2(t)^2}$

• Feature $std(\bar{a}(t))$ where $\bar{a}(t) = \sqrt{a_X(t)^2 + a_Y(t)^2 + a_Z(t)^2}$

WiFi: detects smart phone presence

Access points detect emitted signal: binary 0 or 1



Building ~ 40% of energy usage in the US:

- Plug loads ~ 20% 30% of building energy.
 Benefits of individual presence:
- Improve energy efficiency of demand response, thermal comfort by personalized heating

Results and Discussion



Figure: Misclassification rate after iterations

- Achieves optimal after only
 2 iterations, deteriorates
 afterwards.
- Stopping indicator (marked in red) fires at 2 and frequently during degradation.
- Good indicator for early stopping.



Figure: (left) Ground truth vs. PresenceSense. (right) Learned hourly presence pattern

Figure: Placement of the ultrasonic sensor and acceleration sensor

Figure: Network configuration of ZigBee for ultrasonic sensors

PresenceSense Algorithm

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Algorithm 1 Pseudo-code of PresenceSense
 1: function PRESENCESENSE_LABELING(X, Prior, MaxIter)
  2:
         Initialization:
          L_1^0, L_2^0 \leftarrow Prior(X)
  3:
          stopCond \leftarrow false
  5:
          k \leftarrow 0
          \eta_0 \leftarrow 0.5
  6:
  7:
         Main program:
          while \neg stopCond \land k < MaxIter do
  8:
 9:
                YMat \leftarrow emptyMatrix(n, v)
                for v_{ind} \in \{1, ..., v\} do
10:
                    EstModl \leftarrow ModelEstimation(X^{vind}, L_1^k, L_2^k)
11:
                    YMat(\cdot, v_{ind}) \leftarrow ModelPredict(EstModl, X^{\tilde{v}ind})
12:
                for s_{ind} \in \{1, ..., n\} do
13:
                    Y(s_{ind}) \leftarrow MajorityVote(YMat(s_{ind}, \cdot))
14:
               L_{i,n}^k \leftarrow getSet(Y)
15:
                L_i^{k+1} = \{L_i^k \cap L_{i,n}^k\} \cup Sample\{L_i^k \Delta L_{i,n}^k; \alpha_i\}
16:
               \eta_{k+1} \leftarrow EstEta(L_{1,2}^k, L_{1,2}^{k+1}, \alpha_1, \alpha_2, V^{est})(*)
17:
               stopCond \leftarrow checkStop(L_{1,2}^k, L_{1,2}^{k+1}, \eta_{k,k+1})(*)
18:
               (Optional) search \alpha_1, \alpha_2 such that stopCond \leftarrow false
19:
      by repeating (*) steps
20:
                k \leftarrow k+1
        Outputs: L_1, L_2 \leftarrow L_1^k, L_2^k
21:
```

Idea: Start from common knowledge, iteratively relabels data based on power features Procedure:

 (L:3) Initialize with presence during office hours (8am–6pm) and the rest with absence
 (L:11-14) Train and relabel

by all power classifiers based on the majority rule

3. (L:15-16) Update labeled set by the rule which guarantees convergence (Corollary 1)

4. (L:17-20) Checks stopping condition and reiterate

For more details regarding convergence and stppping rule, please refer to:

Jin, Ming, et al. "PresenceSense: zero-training algorithm for individual presence detection based on power monitoring." BuildSys'14

- User might leave desktop ON when absent.
- Power change and std. dev. (ripple effect) are useful features.
- The presence profile is correctly captured by PresenceSense.

id	Chg/Th	Abs/Th	Prc/Th	PSense
8	<u>.87</u> /.36/.72	.97 / <u>.62</u> / <u>.86</u>	.79/.45/.69	.97/.71/.89
17	.69/.64/.68	.92 / <u>.76</u> / <u>.86</u>	.83/.49/.71	.92/.77/.87
20	.67/.67/.67	.94/.69/ <u>.86</u>	<u>.87</u> /.40/.72	.94/ <u>.68</u> /.87
26	.80/.14/.62	.99 / <u>.66</u> / <u>.90</u>	.87/.15/.67	<u>.96</u> /.84/.93

Table: Comparison ofaccuracy with other popularmodels

PresenceSense outforms
 in most cases (2nd place power
 threshold model).

★Entry **a/b/c**: detection rate when user is absent/present/overall.

Conclusion and Future Work

- Evaluated several presence detection methods: ultrasonic sensor, acceleration sensor, WiFi
- PresenceSense: non-intrusive based on individual power consumption
- Extension to occupancy estimation by sensor fusion
- Adapt to other problems such as indoor positioning without training labels



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