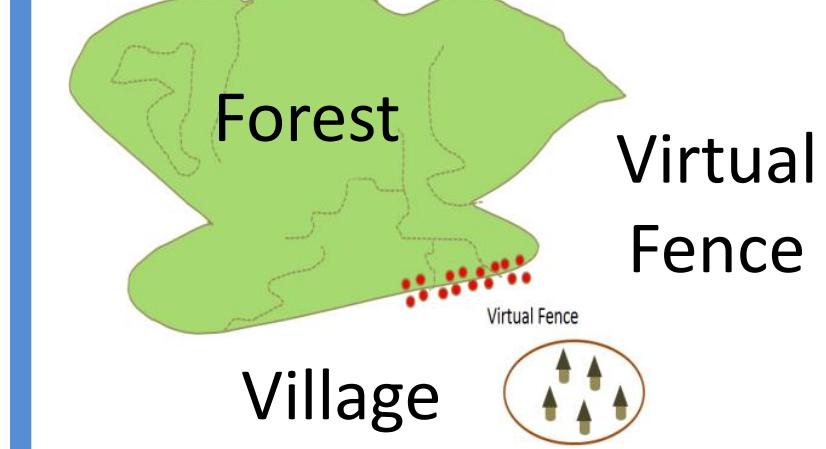


Mitigation of Human-Wildlife Conflicts

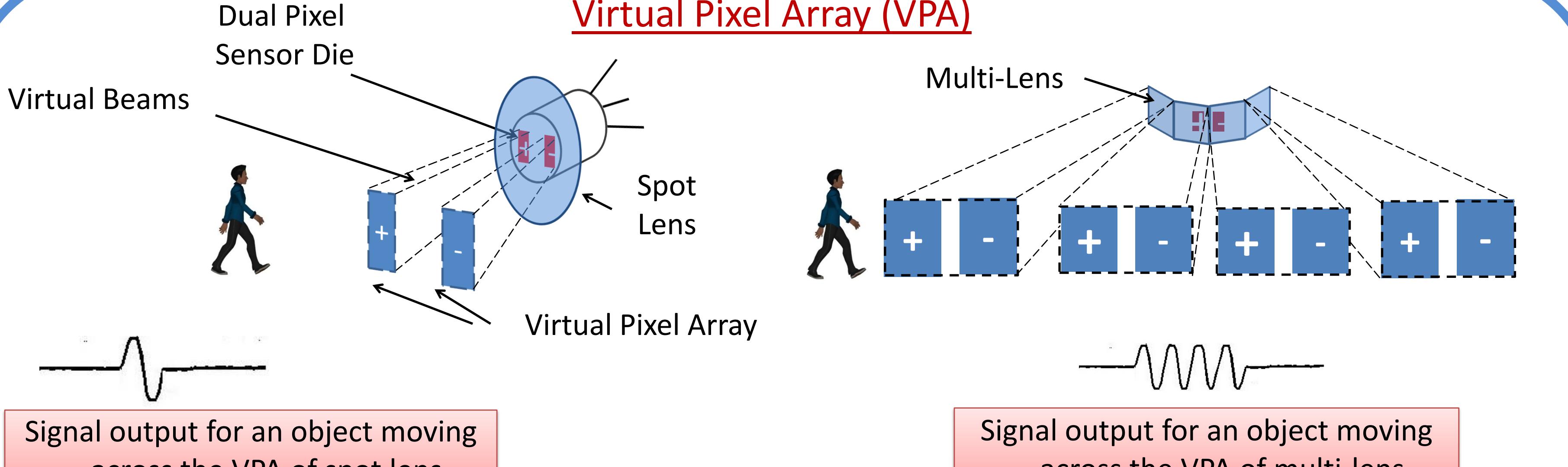




- Animal excursions (killing of livestock and destruction of crops)
- Human intrusions (poaching and forest destruction)

Goal: Investigate efficacy of WSN-based early warning systems to manage human-animal conflicts

Virtual Pixel Array (VPA)



Field of View (FoV): Set of virtual beams along which radiation is received
 VPA associated with a plane: Intersection of FoV with the plane
 Signal generated when an object enters and exits the pixels

Signal output for an object moving across the VPA of spot lens

Signal output for an object moving across the VPA of multi-lens

PIR-Based Sensor Platform for Intruder Classification

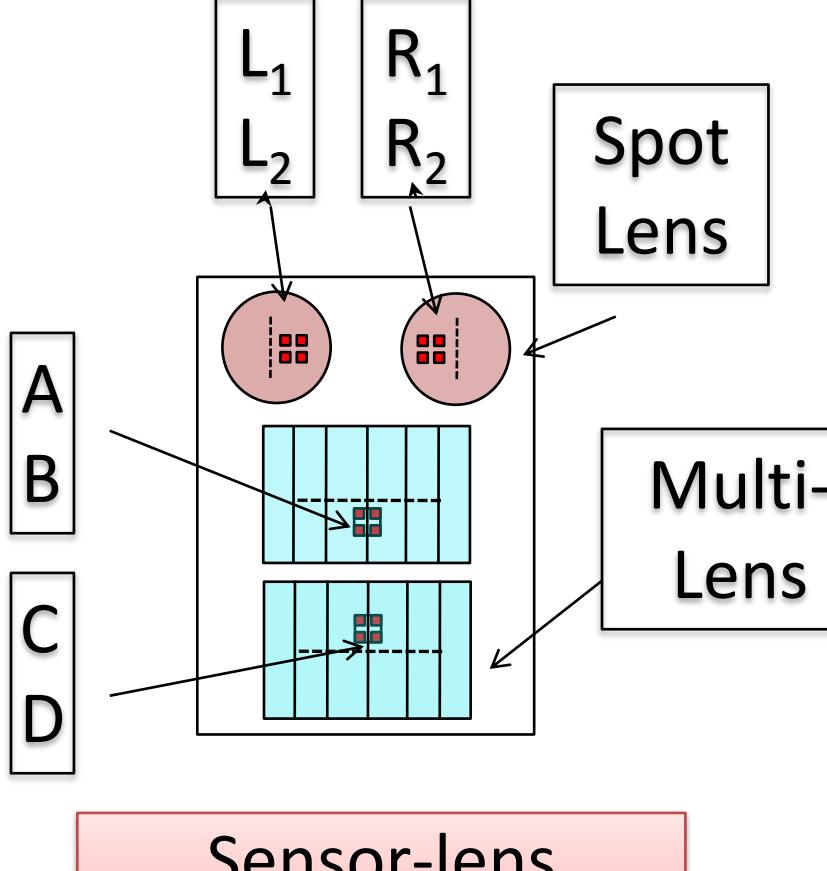


Support Vector Machine Classifier

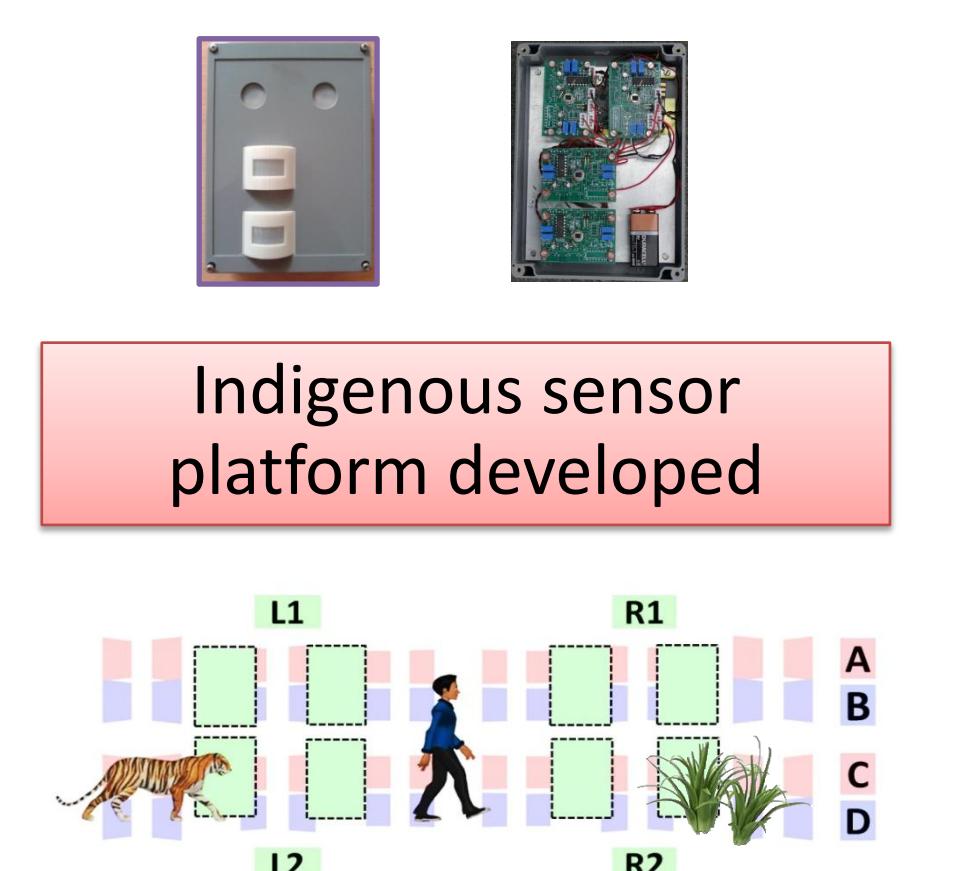


Designed and developed Indigenous sensor platform using inexpensive commercially available components
 Assumptions
 - Intruder moves in straight lines at a uniform velocity
 - No multiple intrusions
 - Only intrusions from humans, dogs, leopards, tigers and wolves

VPA Design

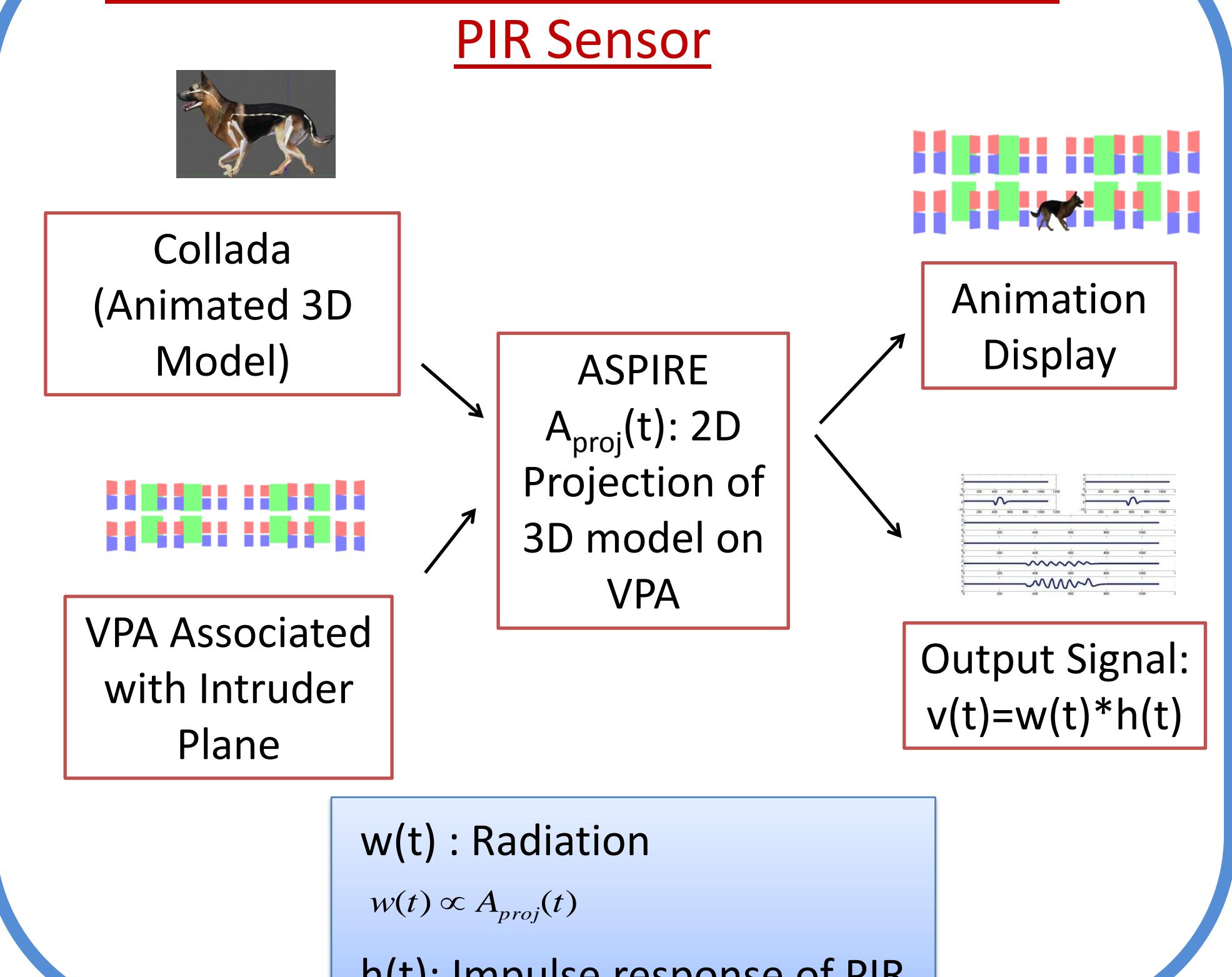


Indigenous sensor platform developed



Waveforms for human, animal and clutter

ASPIRE: Animation-Based Simulation of PIR Sensor



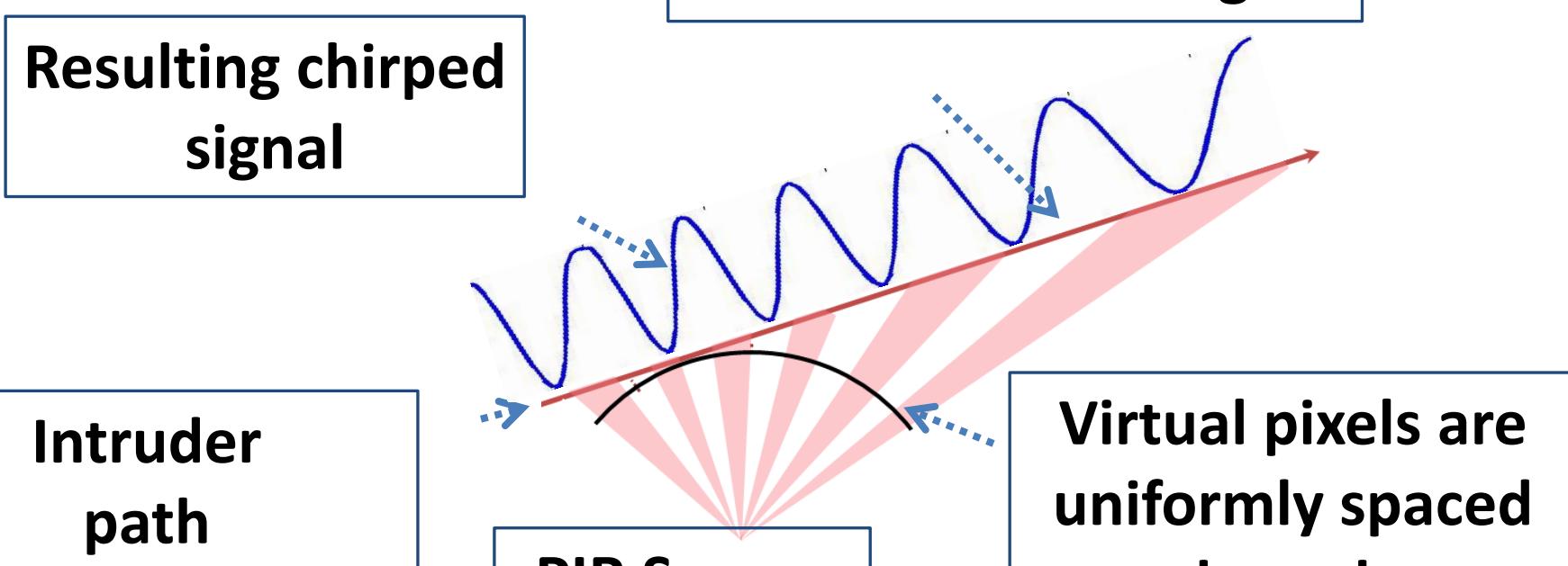
ASPIRE $A_{proj}(t)$: 2D Projection of 3D model on VPA

Output Signal: $v(t) = w(t) * h(t)$

$w(t)$: Radiation
 $w(t) \propto A_{proj}(t)$

$h(t)$: Impulse response of PIR

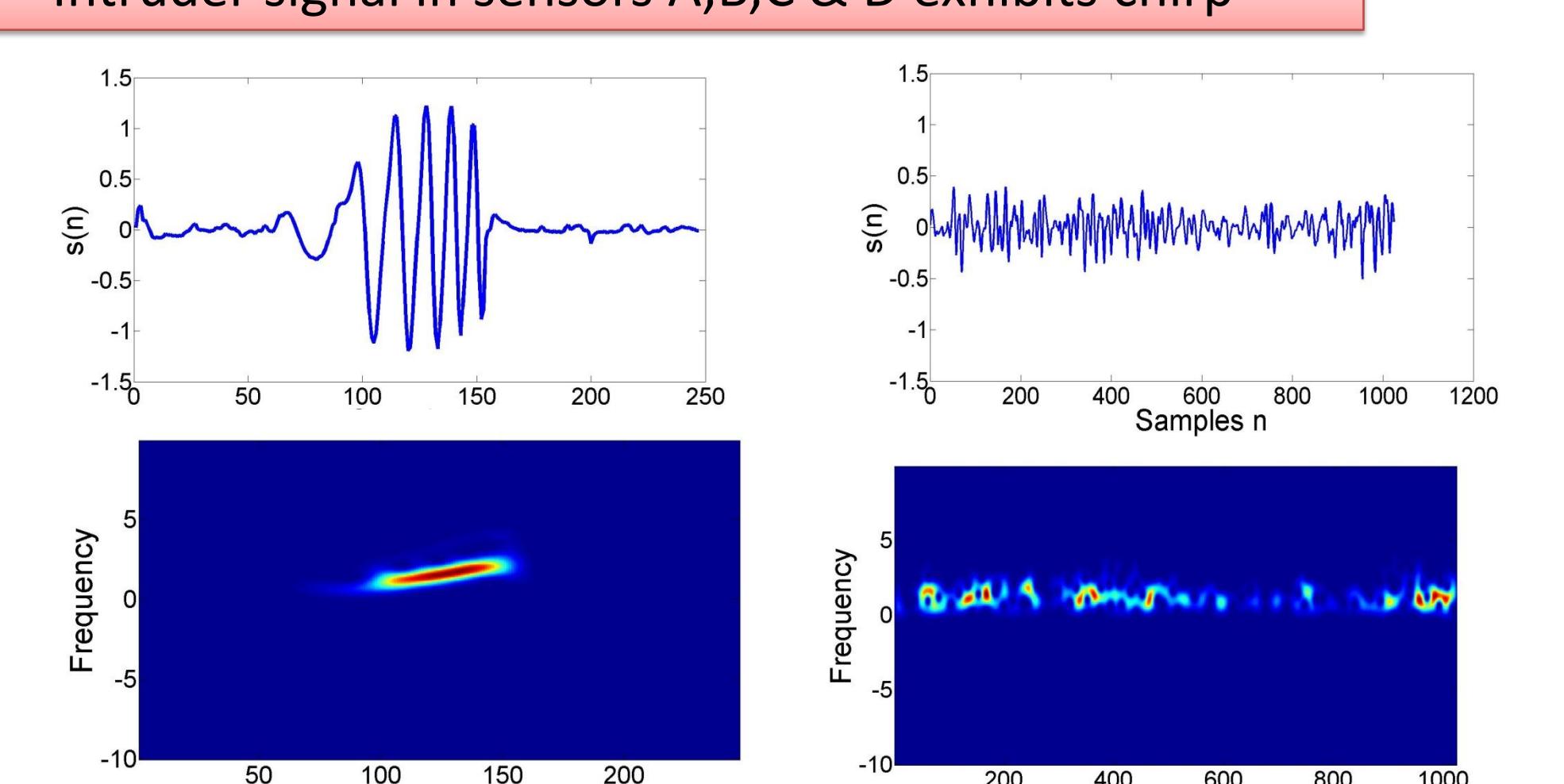
Chirplet-Based Model For Intruder Detection



Spacing between pixels on the intruder path increases to the right

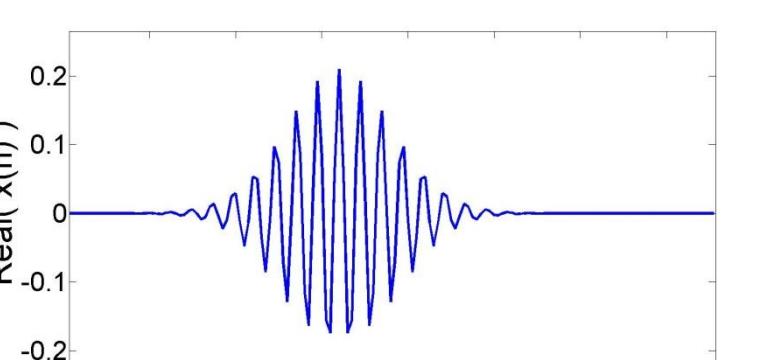
Virtual pixels are uniformly spaced along the circumference

Intruder signal in sensors A, B, C & D exhibits chirp



Intruder and clutter signals & the corresponding STFT

Chirplet-Based Model For Intruder Detection



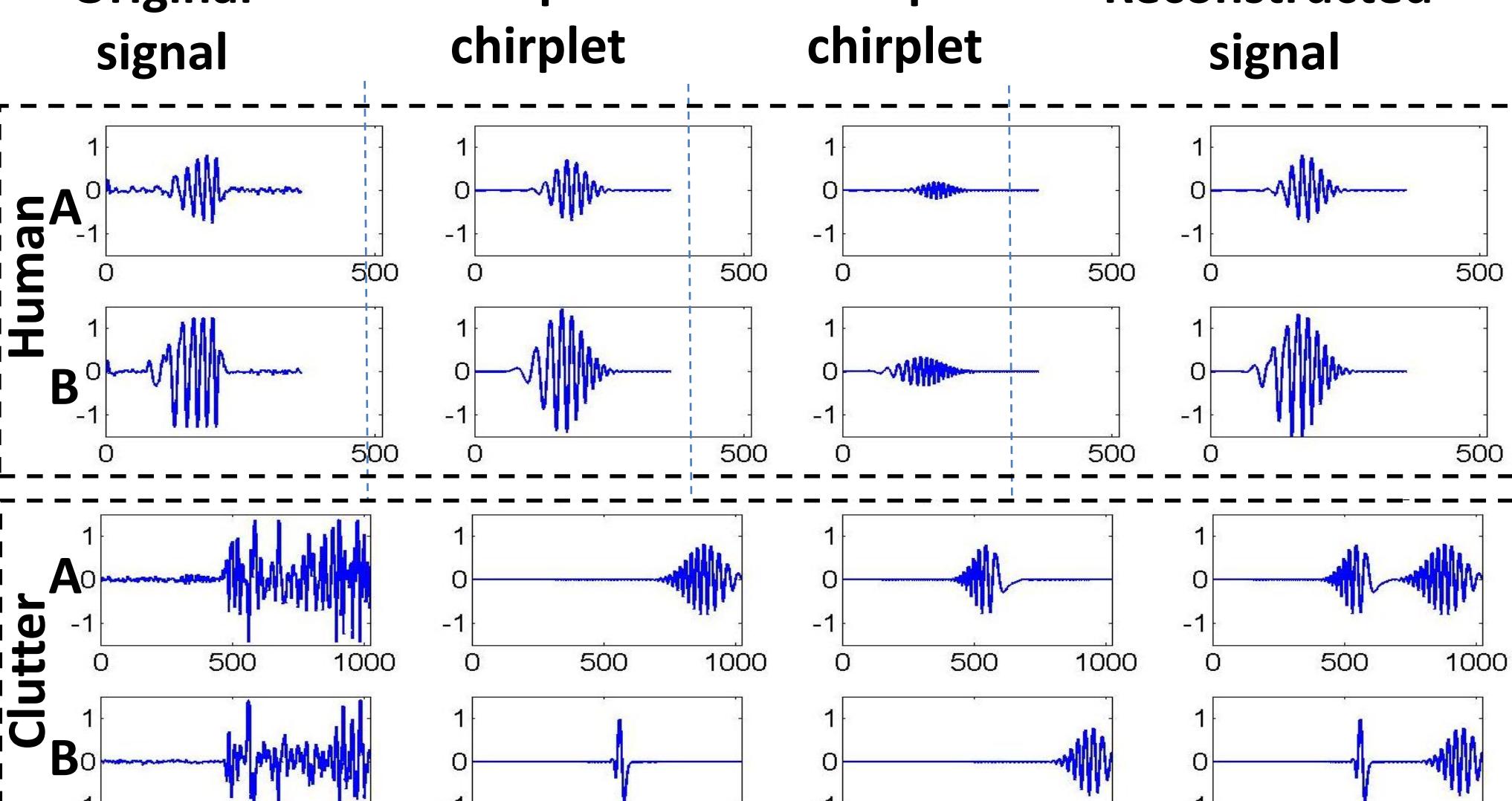
• Chirplet
 $x(n; m, \omega, c, d) = (2\pi d^2)^{-1} \exp\left\{-\frac{(n-m)}{4d^2}\right\} \times \exp\left\{j\frac{c}{2}(n-m)^2 + j\omega(n-m)\right\}$

• Complex analytic representation of signal
 $s_a(n) = s(n) + j\hat{s}(n)$

• Intruder signal well approximated by sum of 3 chirplets:
 $s_a(n) = \sum_{i=1}^3 a_i e^{j\phi} x_i(n; m_i, \omega_i, c_i, d_i)$

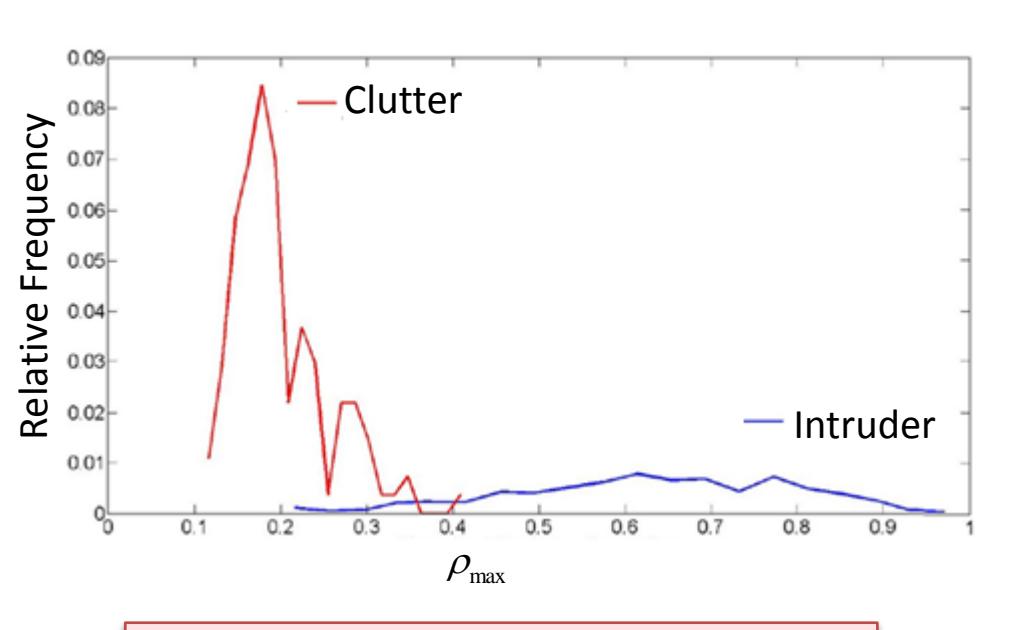
• Chirplet-based feature vector C_{60} : Append ML estimates corresponding to 3 chirplets

Original signal 1st component chirplet 2nd component chirplet Reconstructed signal



Example chirplet decomposition

Energy and Correlation Based Features



Histogram of ρ_{max}

A	B	C	D	Signal Indicates
0	0	0	1	Short animal at 5 m
0	0	1	0	Animal at 10 m
0	0	1	1	Animal at 5 m
0	1	1	0	Human at 10 m
0	1	1	1	Short human at 5 m
1	1	1	1	Tall human at 5 m
(All other combinations)				Clutter or combination unlikely

Inference Drawn from Sensor Combinations Triggered

Feature Performance on Real-World and Simulated Data

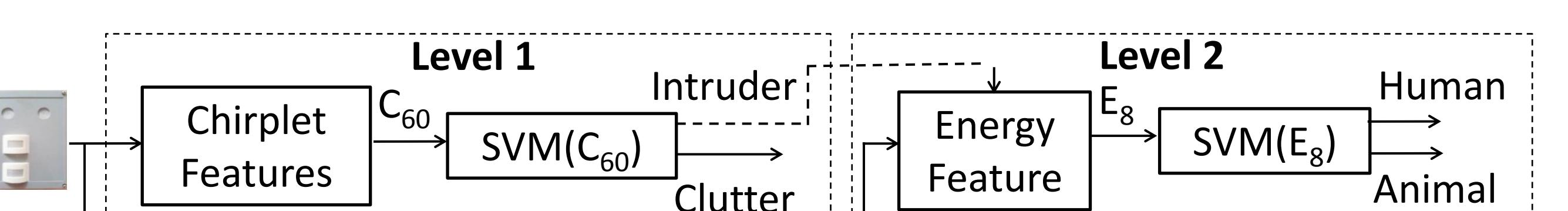
Features	Minimum Accuracy %						Average Accuracy %							
	Intruder			Clutter		Total	Intruder			Clutter		Total		
	E_8	ρ_{max}	C_{60}	RWD	SD	RWD	SD	RWD	SD	RWD	SD	RWD	SD	
✓			94.2	88.6	96.3	87.0	94.8	88.0	97.4	92.9	96.3	92.6	97.1	92.8
✓	✓		95.6	96.2	96.3	87.0	95.8	92.5	97.4	95.0	97.8	94.1	97.5	94.6
	✓	✓	98.6	97.5	100	98.2	99.0	97.8	99.7	99.2	98.5	99.3	99.4	99.3

Intruder vs Clutter

Features	Minimum Accuracy %			Average Accuracy %		
	E_8	Human	Animal	Total	Human	Animal
RWD	95.0	96.7	95.7	98.4	99.3	98.8
SD	100	100	98.8	100	99.5	99.8

Human vs Animal

Final Two-Level Classifier



	Real-World Data		Simulated Data	
	Minimum Accuracy	Average Accuracy	Minimum Accuracy	Average Accuracy
Clutter	96.3	98.3	96.4	99.2
Intruder	100	98.6	98.7	99.2
Human	95.0	98.0	100.0	100.0
Animal	100.0	99.5	100.0	100.0
Overall	98.8	99.9	99.4	99.9

Reference : J. C. O'Neill, P. Flandrin and W. C. Karl, "Sparse representations with chirplets via maximum likelihood Estimation"