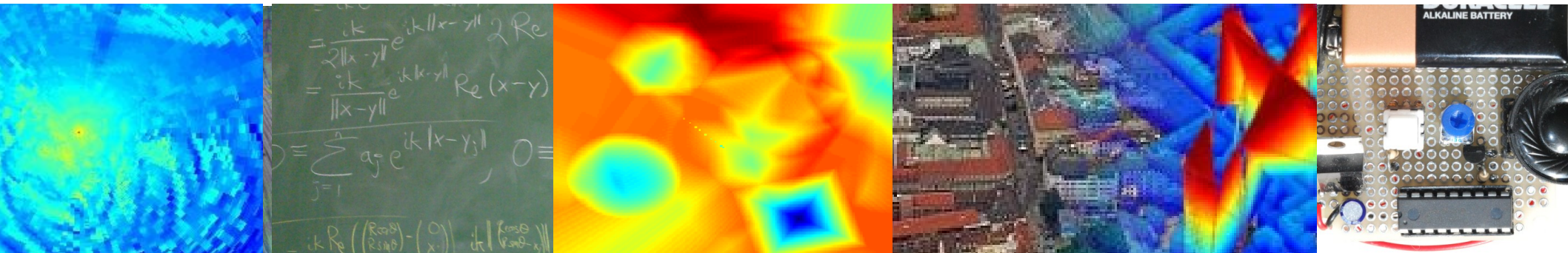


# The Appearance *of* Stratified Spaces *in* Synthetic Aperture Sonar Collections



Michael Robinson



# Acknowledgments

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- Students:
  - Zander Memon (AU)
  - Harry Pham (AU)
  - Maxwell Gualtieri (Northwestern)
- Collaborator:
  - Brian DiZio (NUWC Newport)
- Data: ARL/PSU AirSAS and NUWC HFTAM
- Funding: Kyle Becker (ONR)
- Main references:



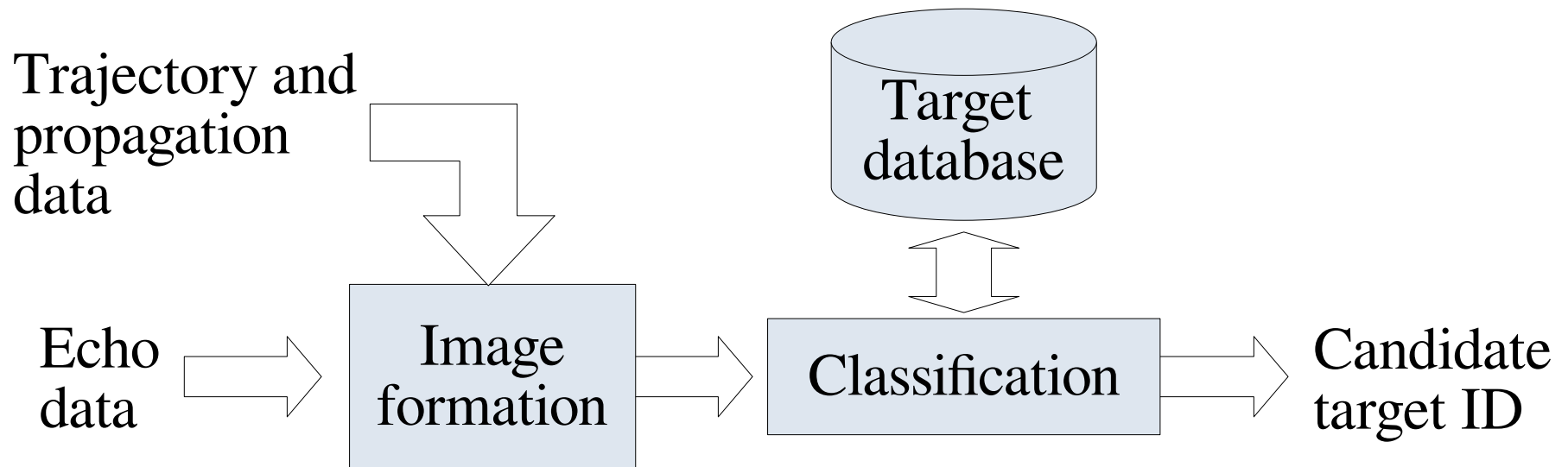
M. Robinson, Z. Memon, M. Gualtieri, “Topological and geometric characterization of synthetic aperture sonar collections.” *Journal Acoustic Society of America*, 2025. <https://doi.org/10.1121/10.0037085>

Z. Memon, M. Robinson, "The Topology of Circular Synthetic Aperture Sonar Targets." [arXiv:2205.11311](https://arxiv.org/abs/2205.11311)



# Motivation

- Goal: Detect and identify objects on the seafloor using their active sonar signatures



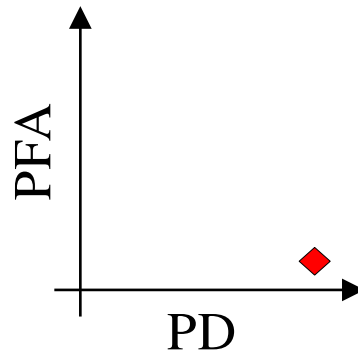
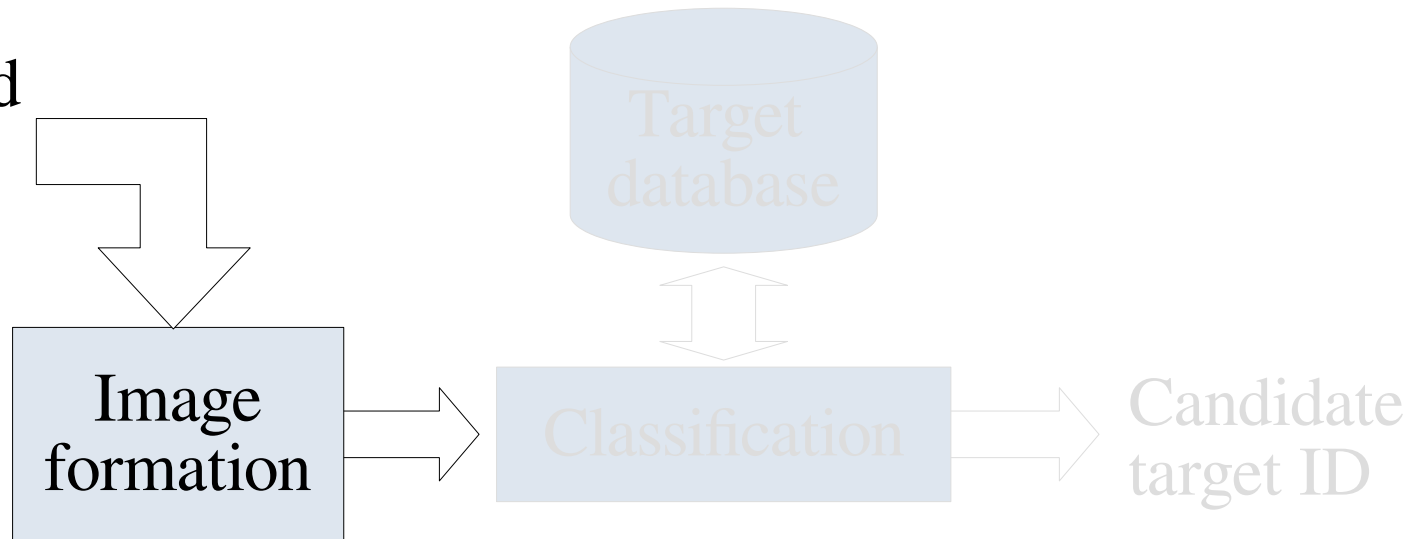
- Nearly all research, development, and fielded systems follow this common pipeline

# Motivation

- Under tightly controlled conditions, sonar image quality is very good

Trajectory and  
propagation  
data

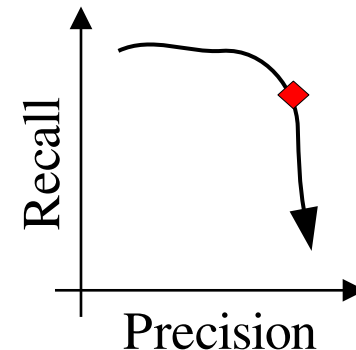
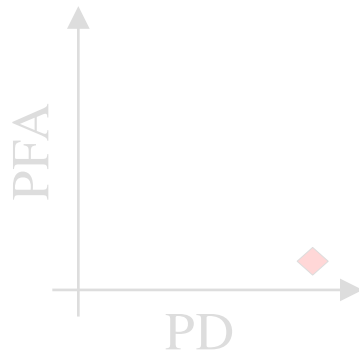
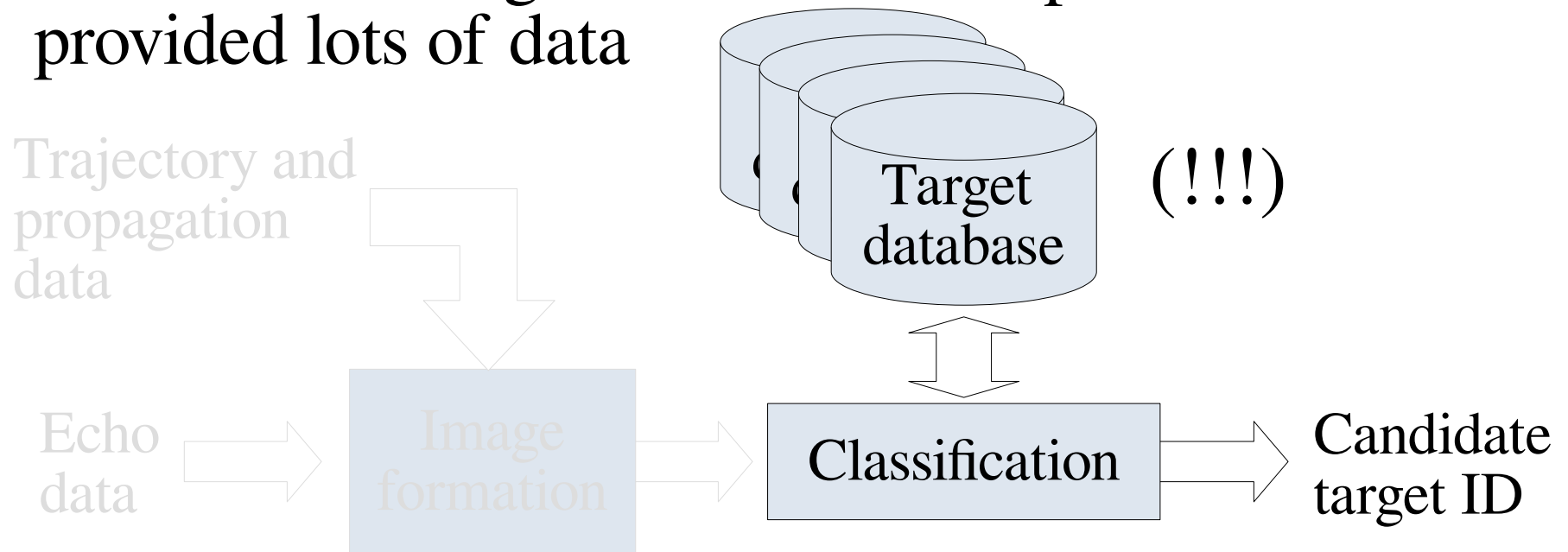
Echo  
data



Target detection performance is good  
Focus, contrast, etc. good too

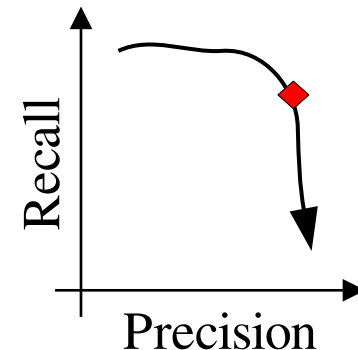
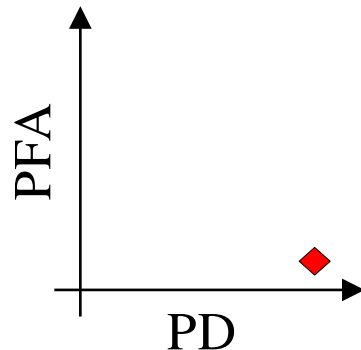
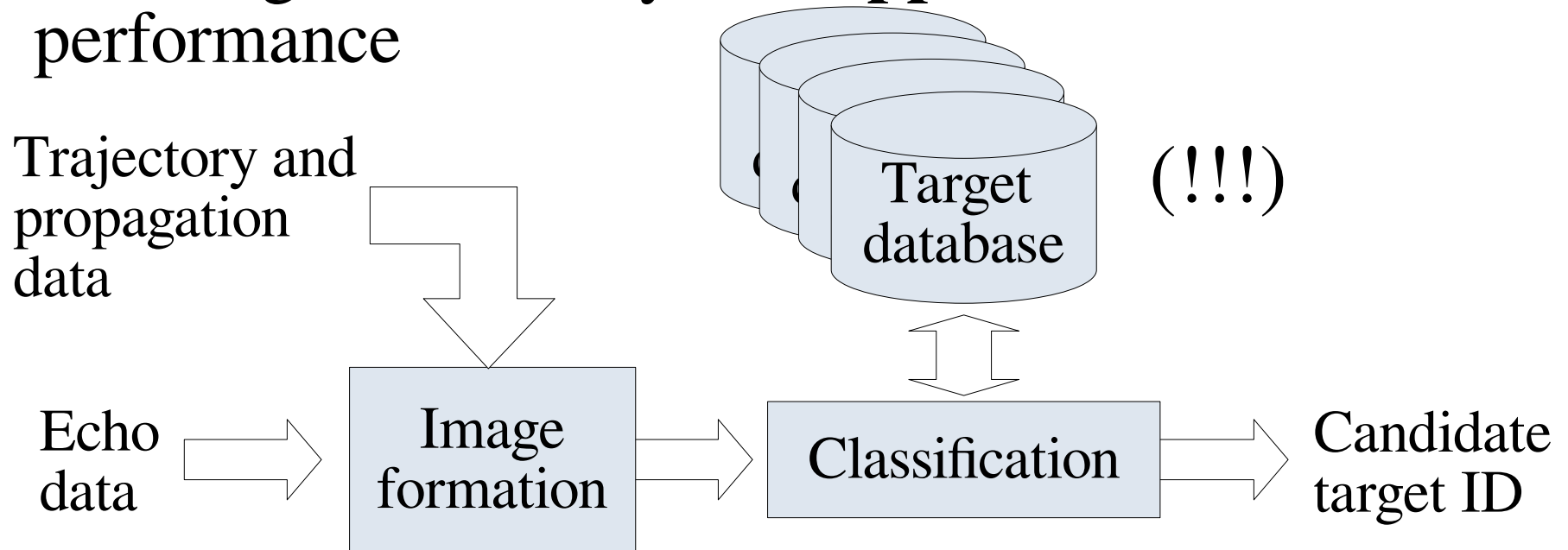
# Motivation

- Machine learning-based classifiers perform well if provided lots of data



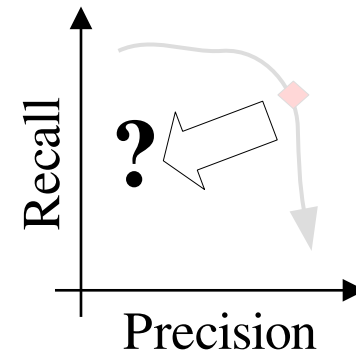
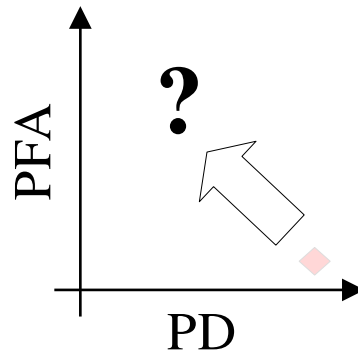
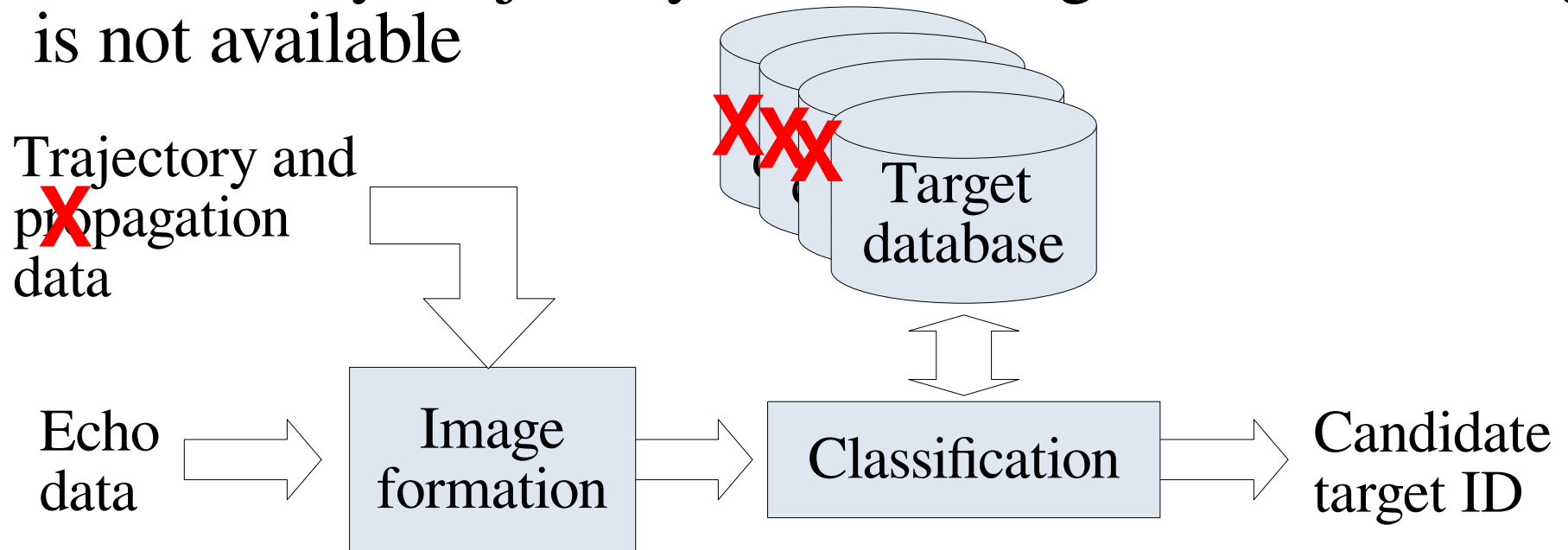
# Motivation

- Chaining these tools yields **upper bound** on overall performance



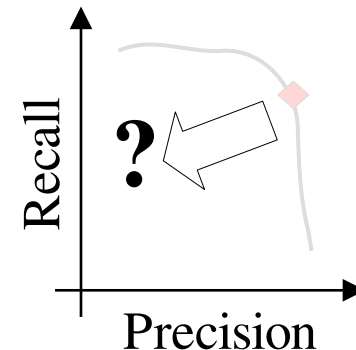
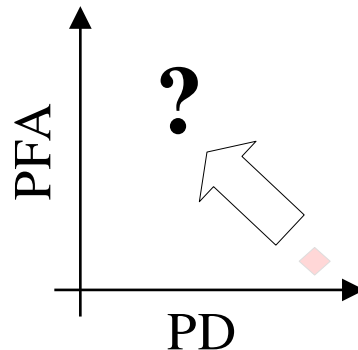
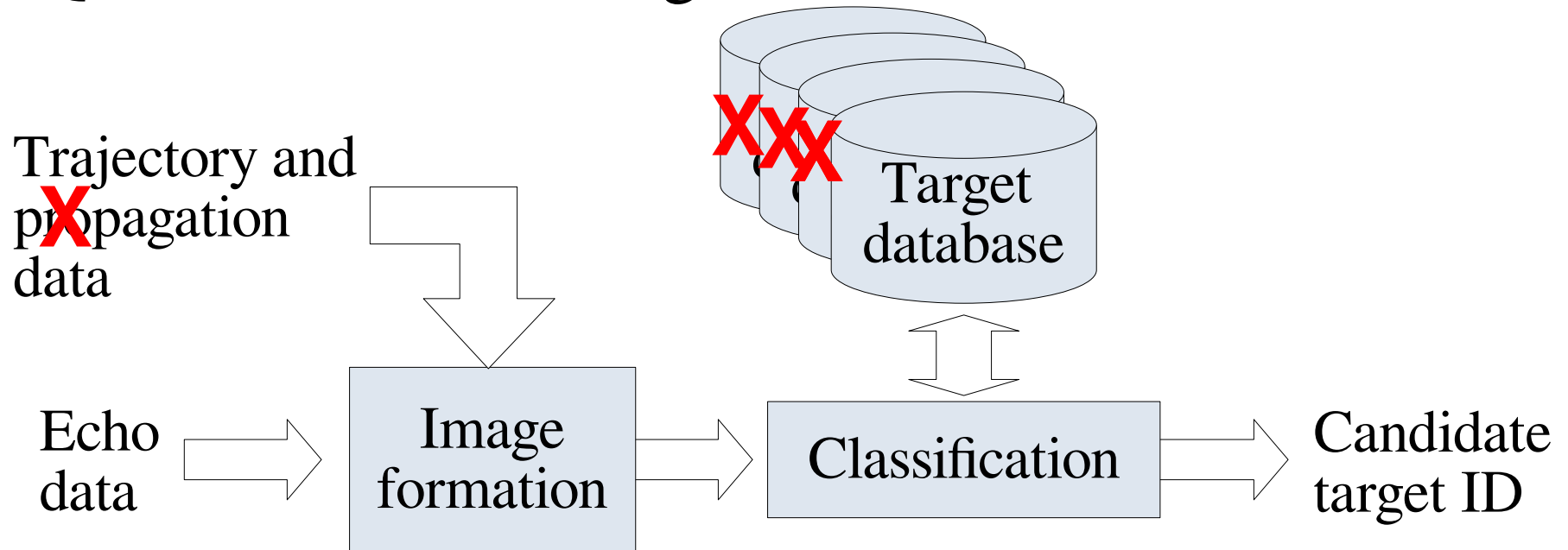
# Motivation

- Realistically, trajectory won't be as good; vast training is not available



# Motivation

- Question: What strategies remain in this case?





# Motivation

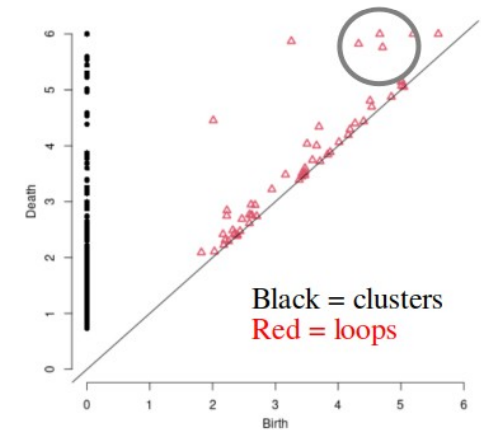
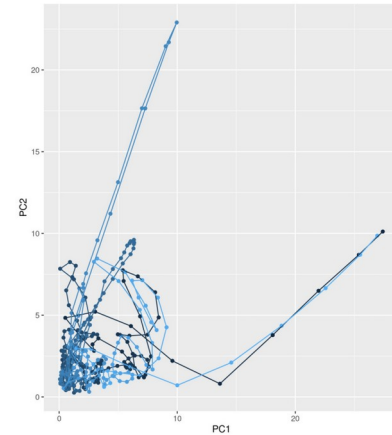
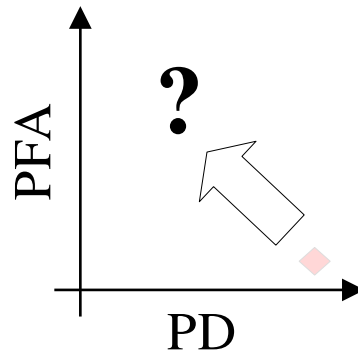
- Answer: Trajectory-invariant topological signal rep'ns

Trajectory and  
~~propagation~~  
data

Echo  
data

Image  
formation

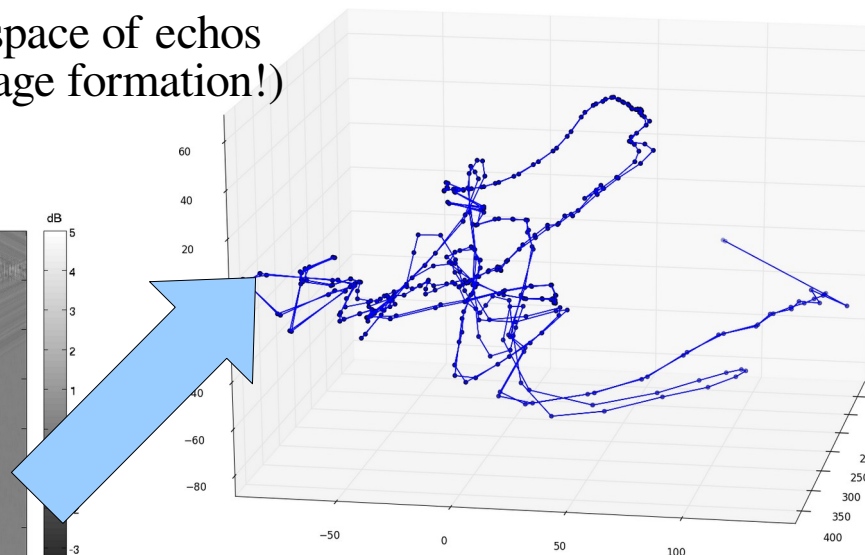
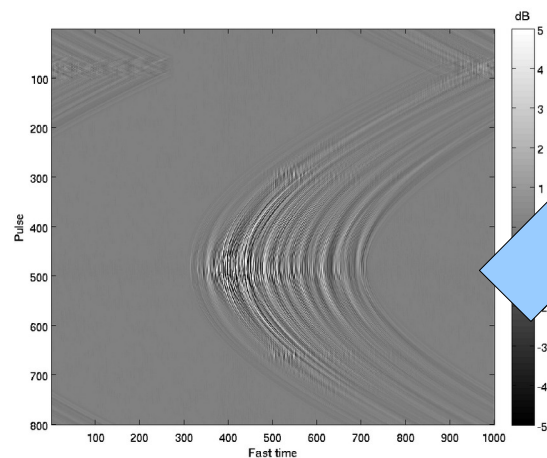
Topological  
signature  
features



# Initial insight: topological tools **do** work

Metric space of echos  
(**not** image formation!)

Pulses



(Processed data is representative only!)

Persistence  
diagram

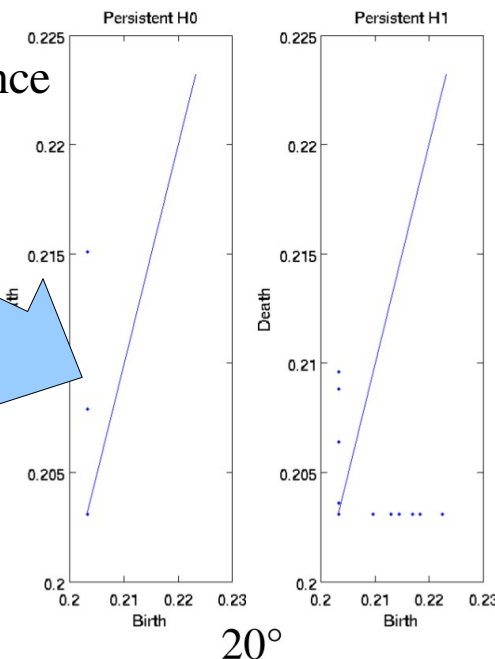


Table of mean  
misclassification  
rates for various  
metrics

(smaller is better)

Metric	Target types	Target groups
Spectral $L^2$	4.32	2.32
Spectral corr.	3.58	2.19
Tucker [6, Fig. 5(b)]	2.15	1.57
$H_0$ with $L^2$	<b>1.47</b>	<b>1.28</b>
$H_1$ with $L^2$	2.49	1.70
$H_0$ with corr.	1.62	1.30
$H_1$ with corr.	2.38	1.57

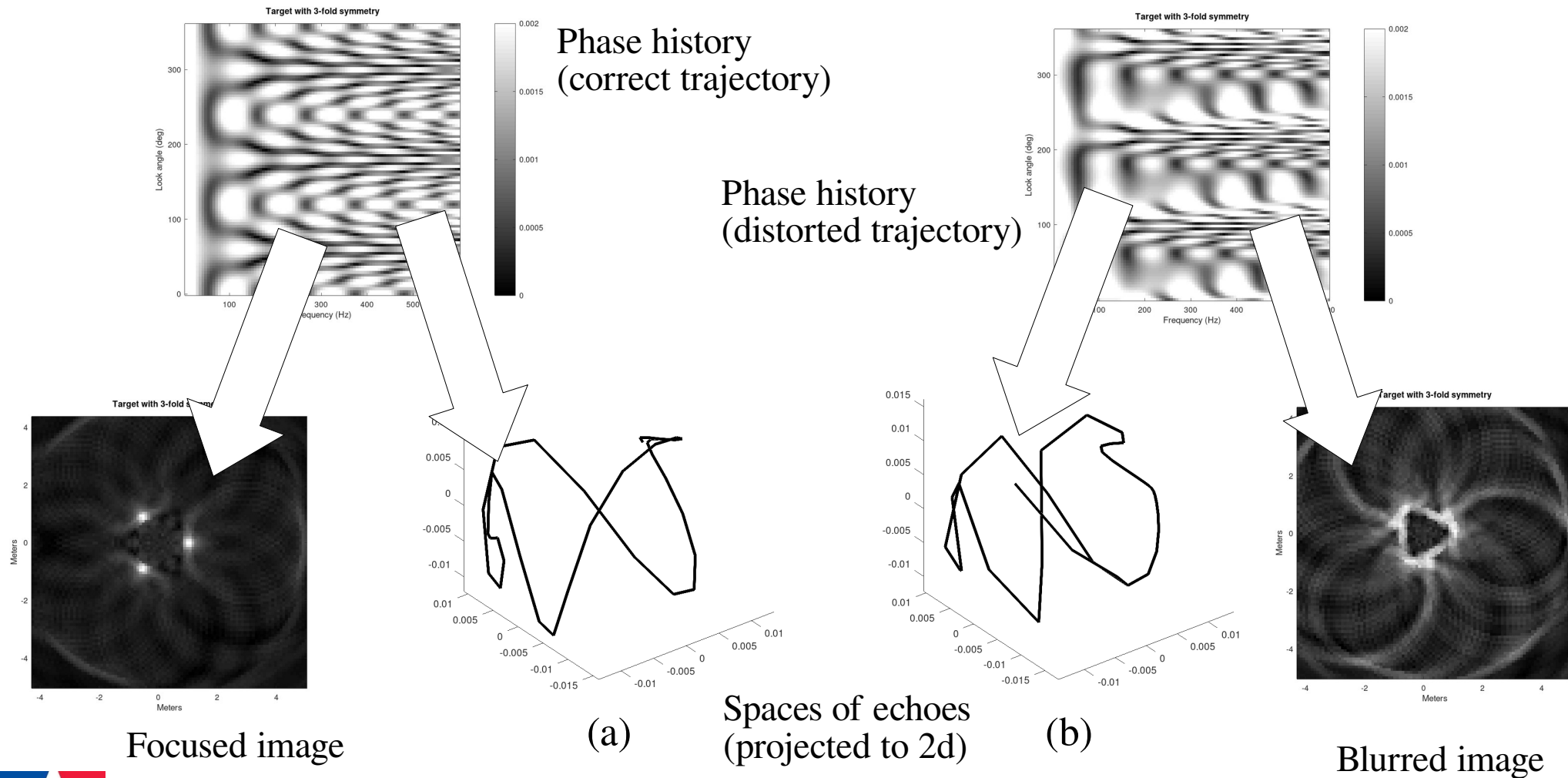
Non-topological

Topological

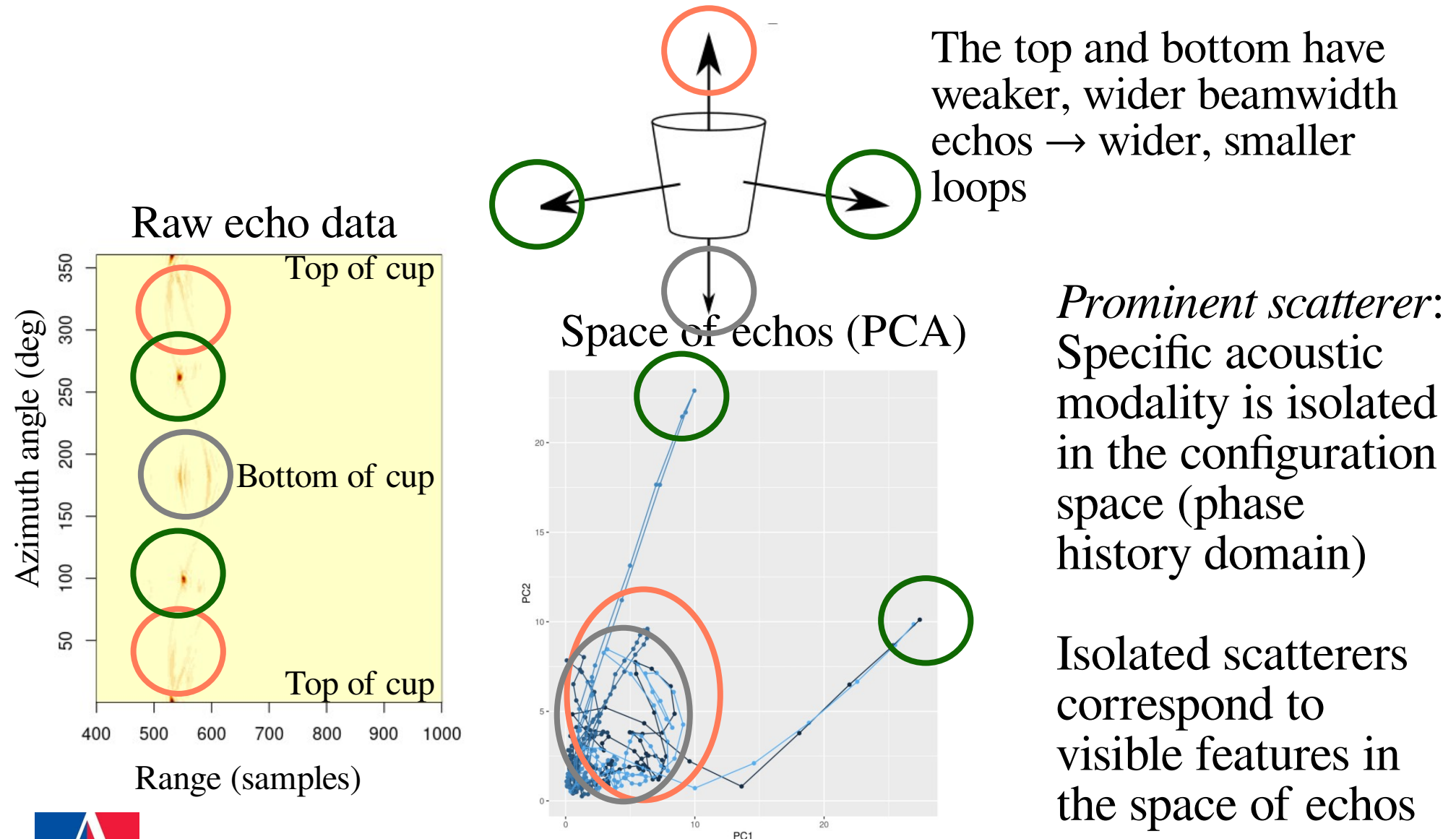


# Topological features are robust

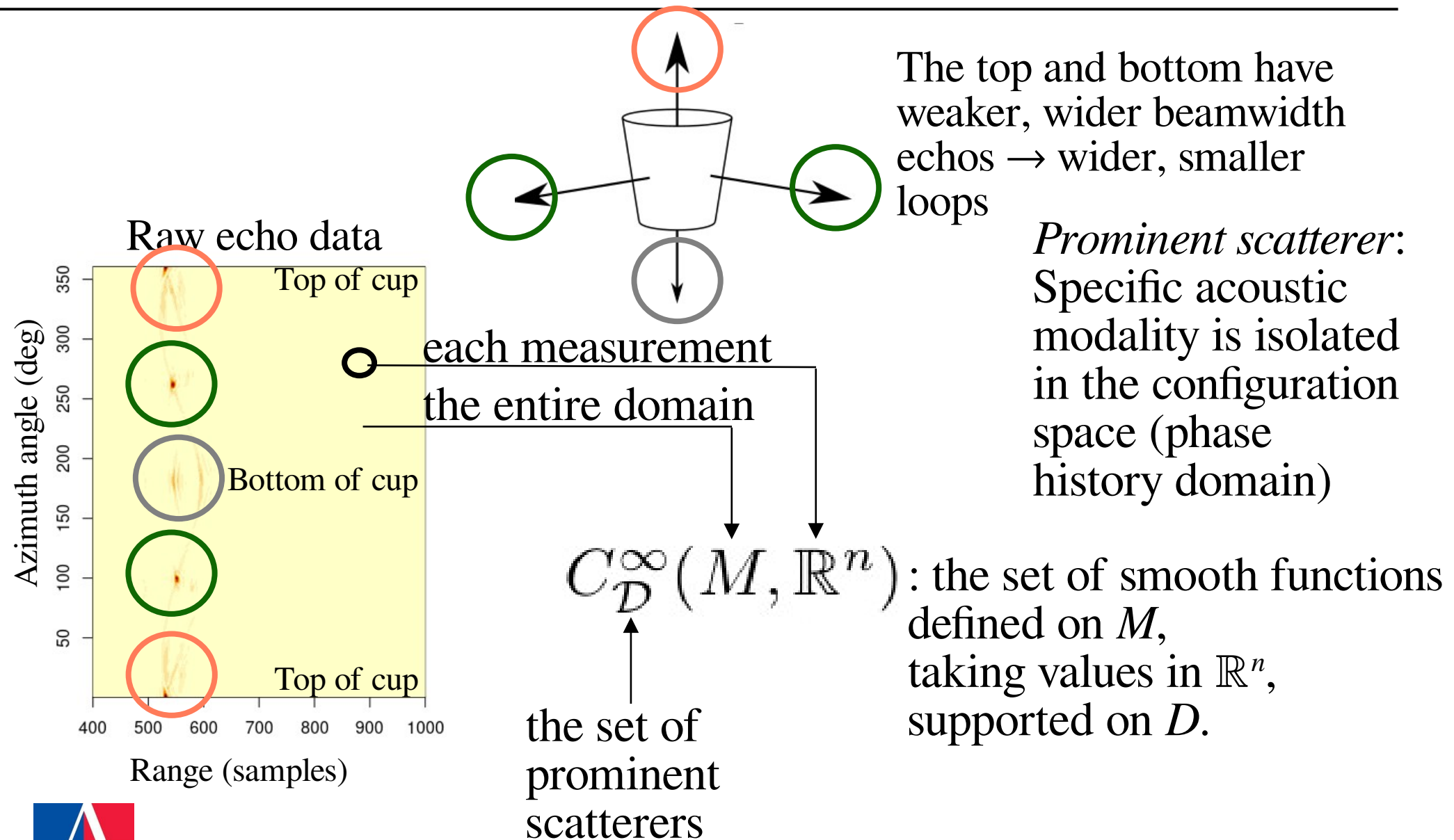
- The space of echoes ignores trajectory distortions



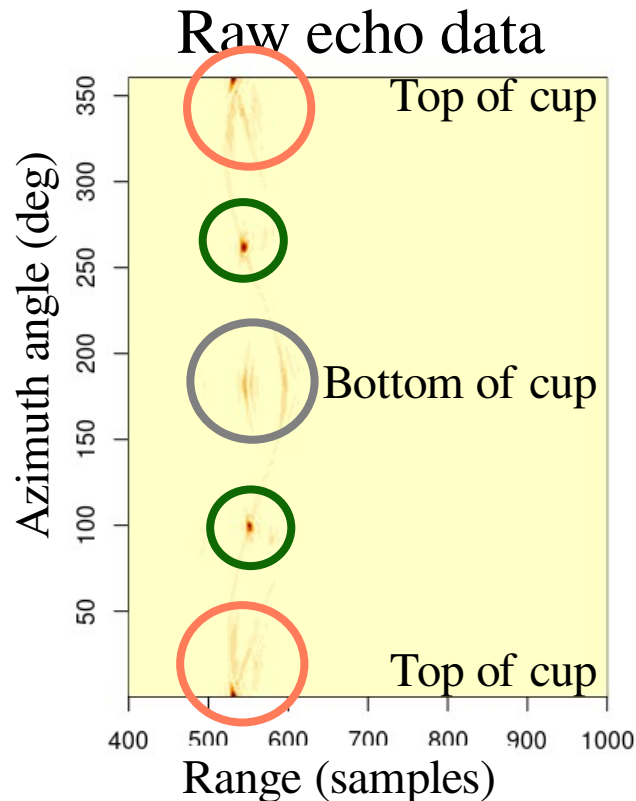
# New insight: Prominent scatterers



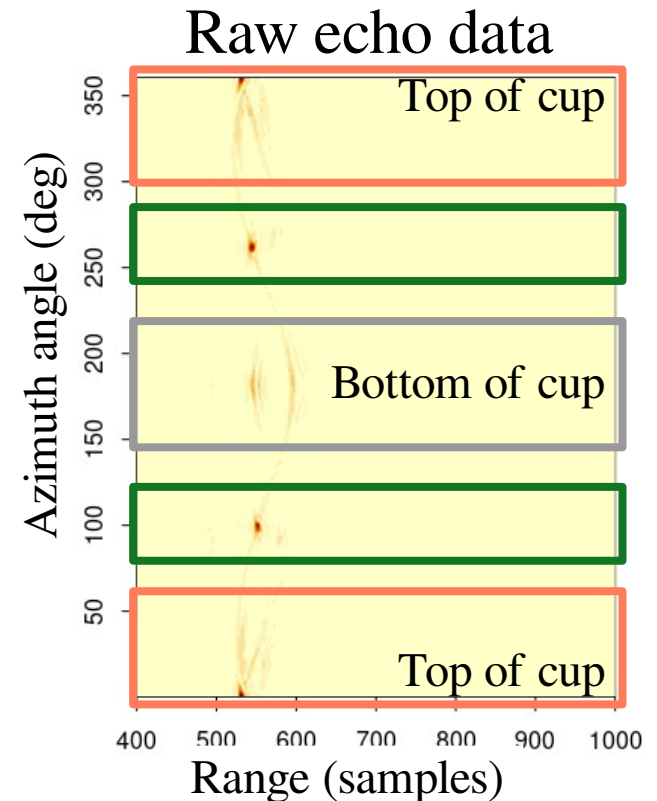
# Topological space of echos



# Flexibility in the signal models



Topology says:  
Both models  
equally valid...  
but one might  
be more useful!



azimuth

range

$$C_D^\infty(S^1 \times \mathbb{R}, \mathbb{R})$$

single sample

azimuth

$$C_D^\infty(S^1, \mathbb{R}^n)$$

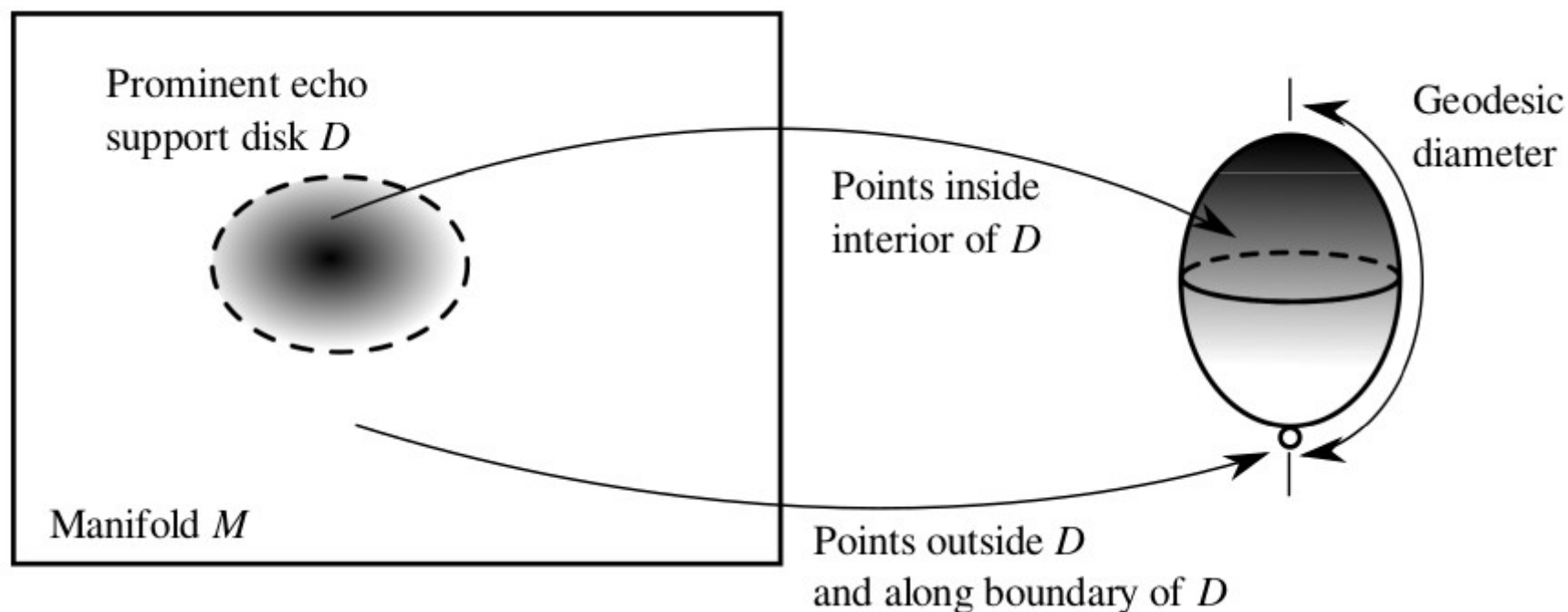
vector of range samples



# Signals to spaces (our main result)

**Theorem 1.** *Let  $M$  be a smooth manifold and suppose that  $n > 2 \dim M$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the signature space of  $v$  is homeomorphic to a wedge product of spheres of (intrinsic) dimension  $\dim M$ .*

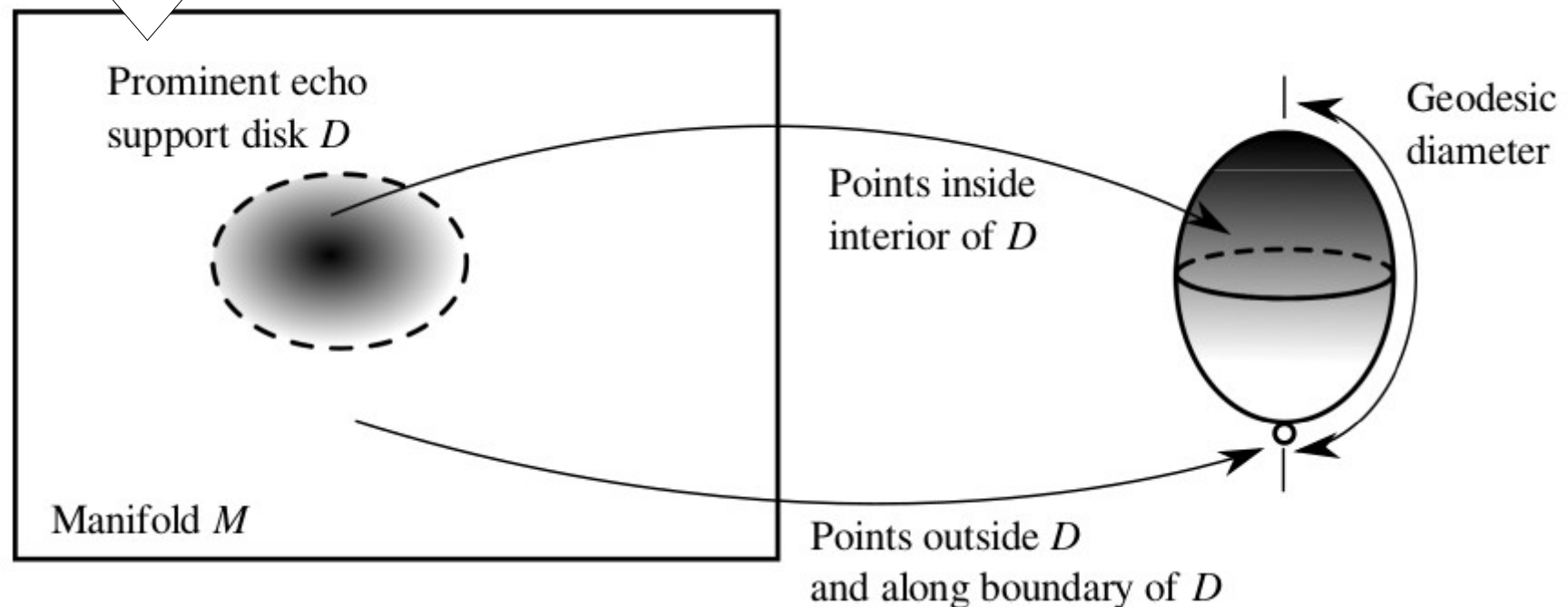
*Moreover, each prominent echo corresponds to a sphere of the same dimension as  $M$  in the signature space. Under the usual metric for  $\mathbb{R}^n$ , the cross section for a prominent echo is a lower bound for the geodesic diameter of its corresponding sphere<sup>3</sup>.*



# A sonar signature is a function on $M$

**Theorem 1.** Let  $M$  be a smooth manifold and suppose that  $n > 2 \dim M$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the signature space of  $v$  is homeomorphic to a wedge product of spheres of (intrinsic) dimension  $\dim M$ .

Moreover, each prominent echo corresponds to a sphere of the same dimension as  $M$  in the signature space. Under the usual metric for  $\mathbb{R}^n$ , the cross section for a prominent echo is a lower bound for the geodesic diameter of its corresponding sphere.

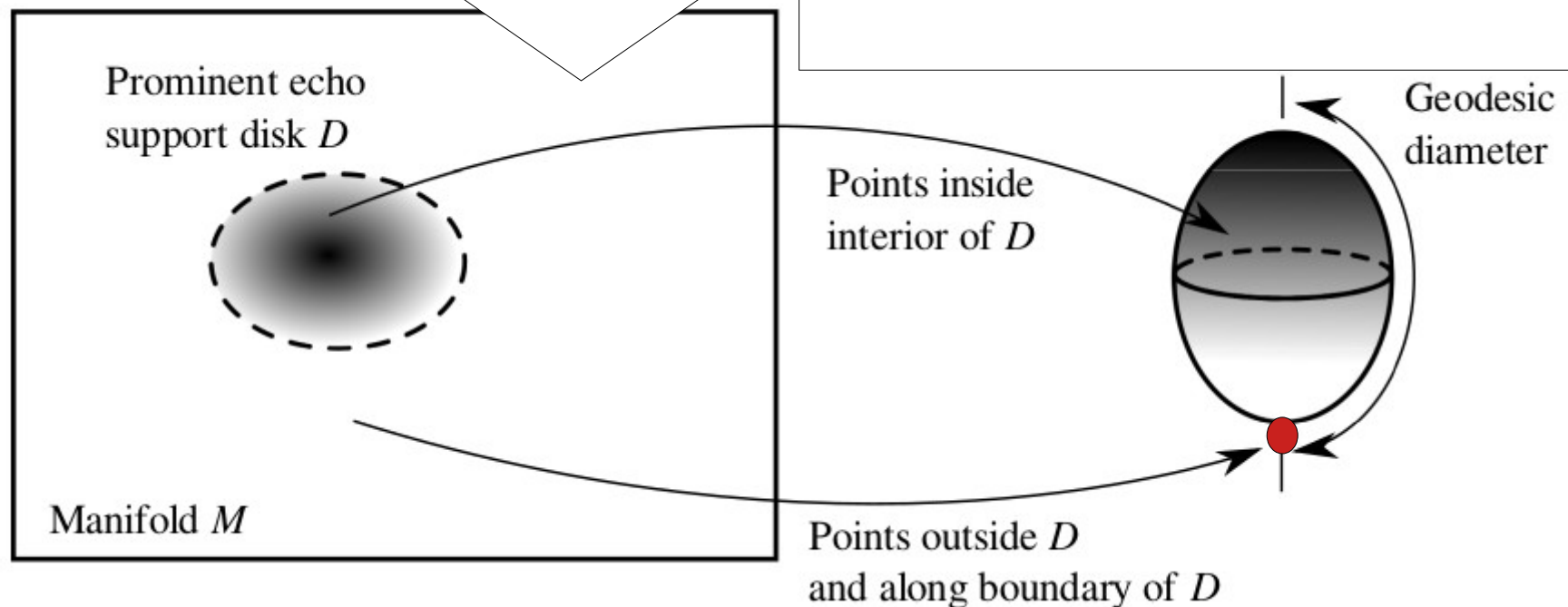




# Its image is a “bouquet” of spheres

**Theorem 1.** *Let  $M$  be a smooth manifold and suppose that  $n > 2 \dim M$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the signature space of  $v$  is homeomorphic to a wedge product of spheres of (intrinsic) dimension  $\dim M$ .*

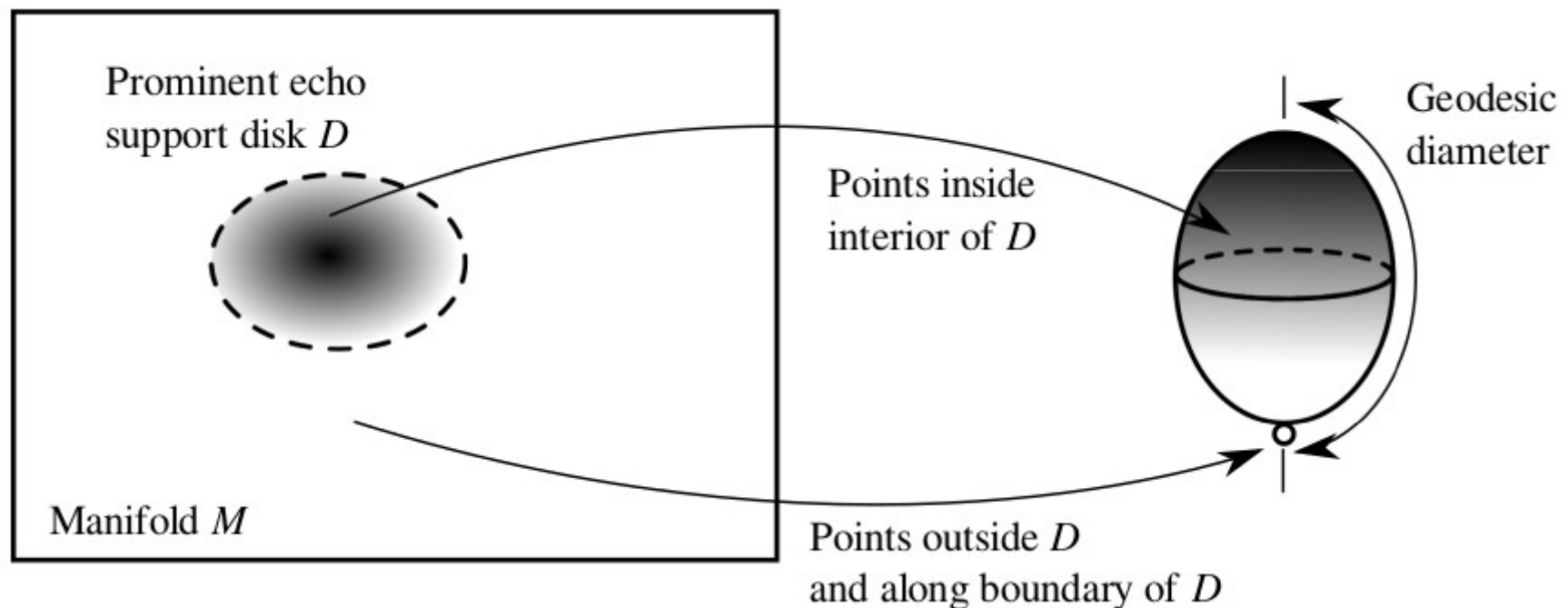
Moreover, each prominent echo corresponds to a sphere of the same dimension as  $M$  in the signature space. Under a generic section for a prominent echo in the lower boundary, every point outside prominent scatterers yields no echo, so it maps to the same value.



...with typically enough room to not overlap

**Theorem 1.** Let  $M$  be a smooth manifold and suppose that  $n > 2 \dim M$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the signature space of  $v$  is homeomorphic to a wedge product of spheres of (intrinsic) dimension  $\dim M$ .

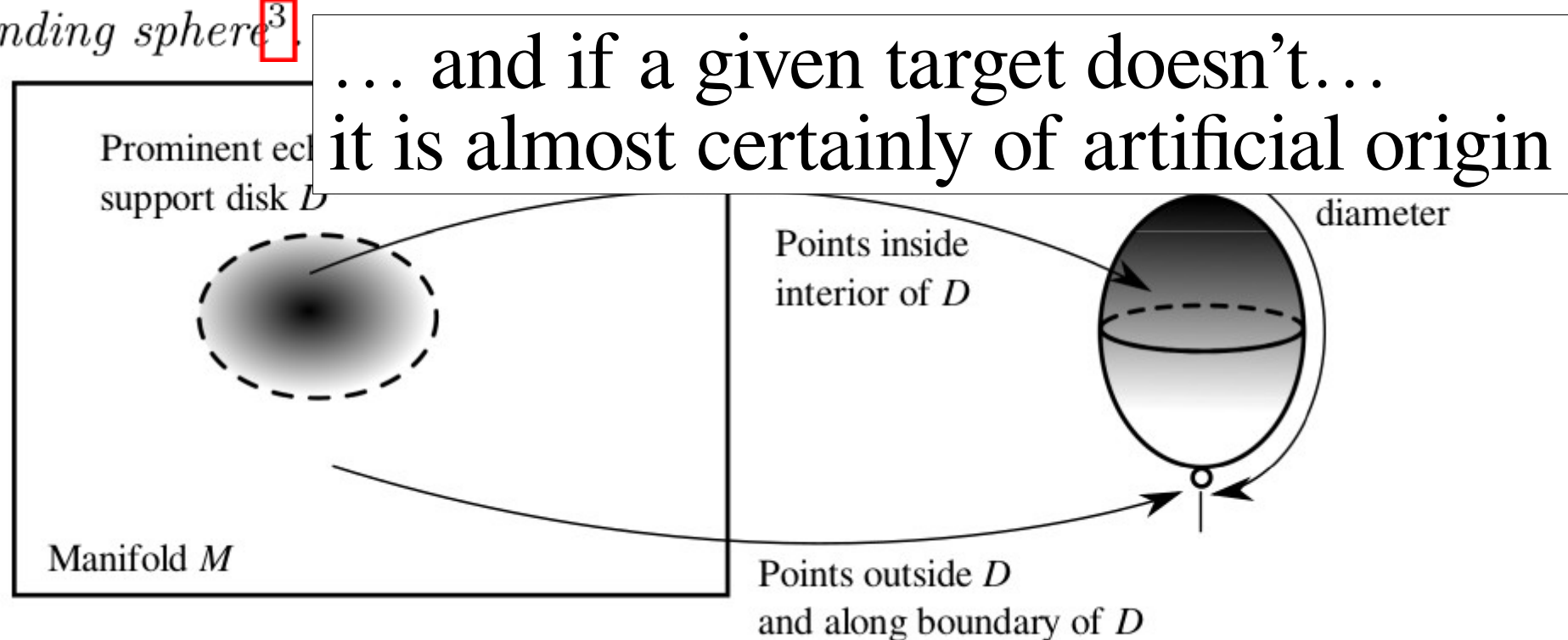
Moreover, each prominent echo corresponds to a sphere of the same dimension as  $M$  in the signature space. Under the usual metric for  $\mathbb{R}^n$ , the cross section for a prominent echo is a lower bound for the geodesic diameter of its corresponding sphere<sup>3</sup>.



# Typical targets will look like this

**Theorem 1.** *Let  $M$  be a smooth manifold and suppose that  $n > 2 \dim M$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the signature space of  $v$  is homeomorphic to a wedge product of spheres of (intrinsic) dimension  $\dim M$ .*

Moreover, each prominent echo corresponds to a sphere of the same dimension as  $M$  in the signature space. Under the usual metric for  $\mathbb{R}^n$ , the cross section for a prominent echo is a lower bound for the geodesic diameter of its corresponding sphere<sup>3</sup>.



# Testing this as a null hypothesis

---

- Homeomorphisms cannot be tested directly
- But a weaker test is homology, and that we can compute directly from the theorem:

Estimable via *persistent homology* →

$$\dim H_k(v(M)) = \begin{cases} 1 & \text{if } k = 0, \\ \# \mathcal{D} & \text{if } k = \dim M, \\ 0 & \text{otherwise.} \end{cases}$$

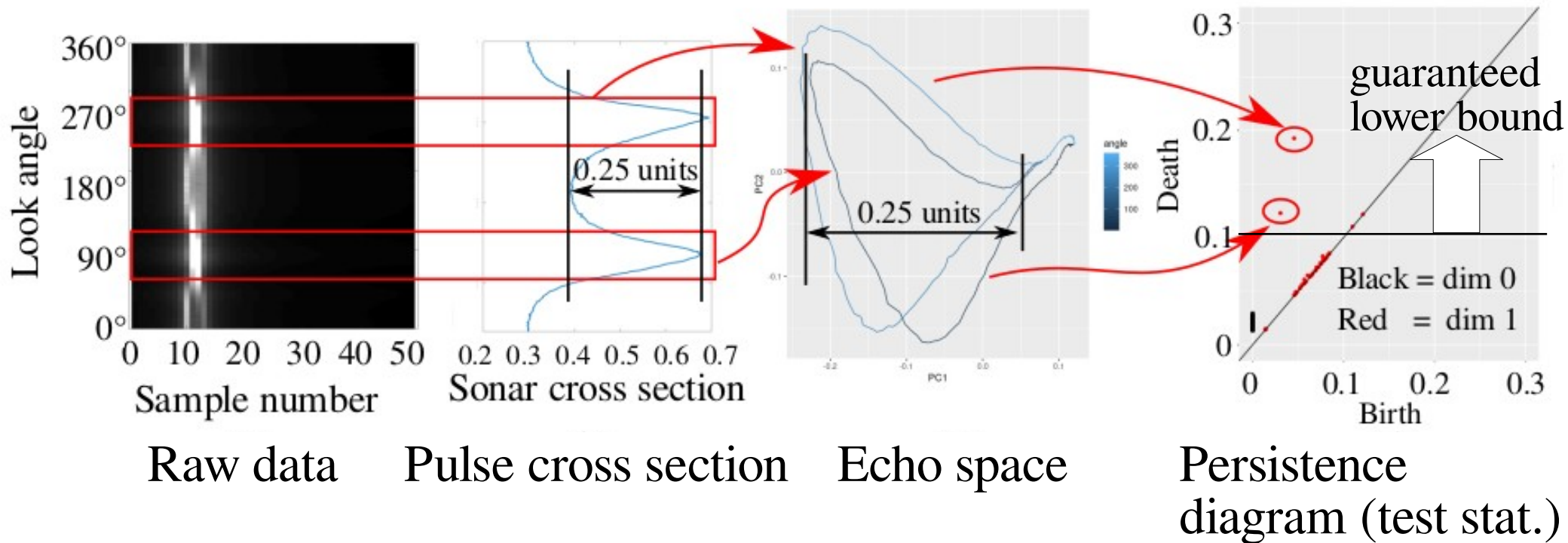
← measurable! (pointing to the case  $k=0$ )

← known! (pointing to the case  $k = \dim M$ )

- Under noisy conditions, persistent homology is a test statistic\* for “target unusualness”
- Takeaway: We **finally** have theoretical justification for using topology to classify!

\*Omer Bobrowski and Primoz Skraba. A universal null-distribution for topological data analysis. *Scientific Reports*, 13(1):12274, 2023.

# Simulation check: point scatterers

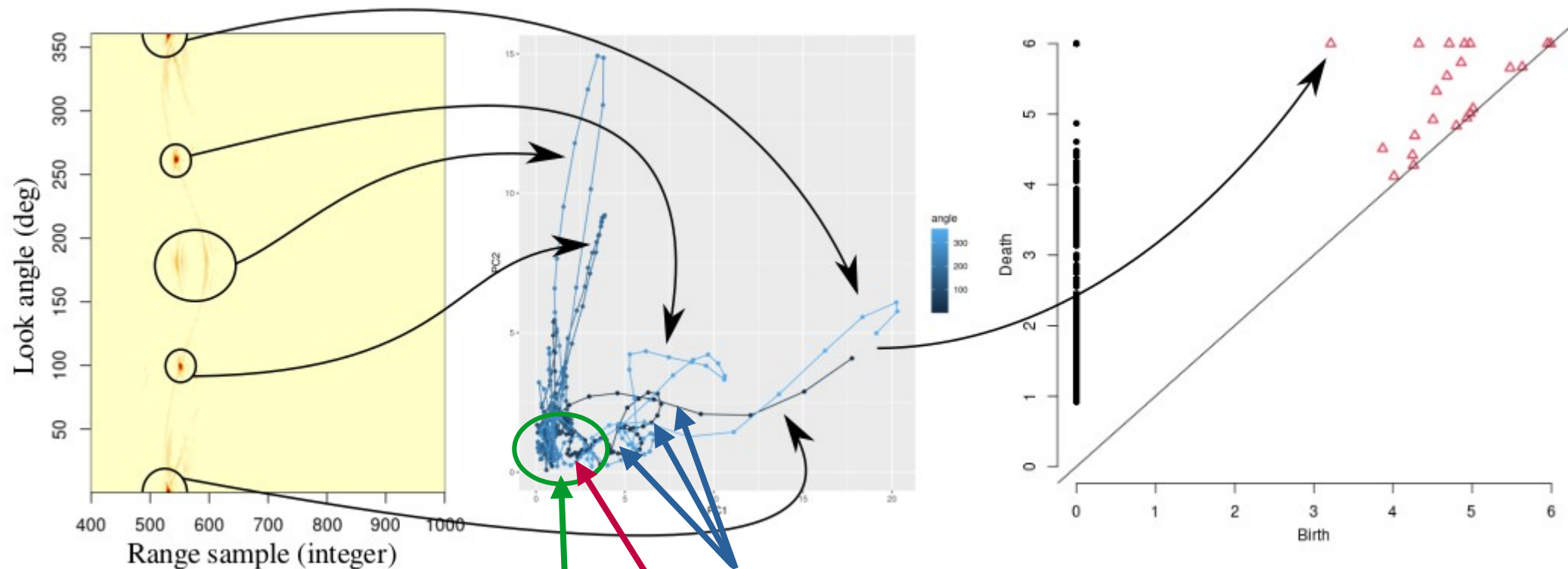


**Proposition 7.** *For the 1-dimensional embedded Vietoris-Rips filtration  $VR_\epsilon$  constructed from the signature space as above, each prominent echo corresponds to a generator with death time bounded below by  $\sigma/2$ .*

$0.25 / 2 = 0.125$  in this case



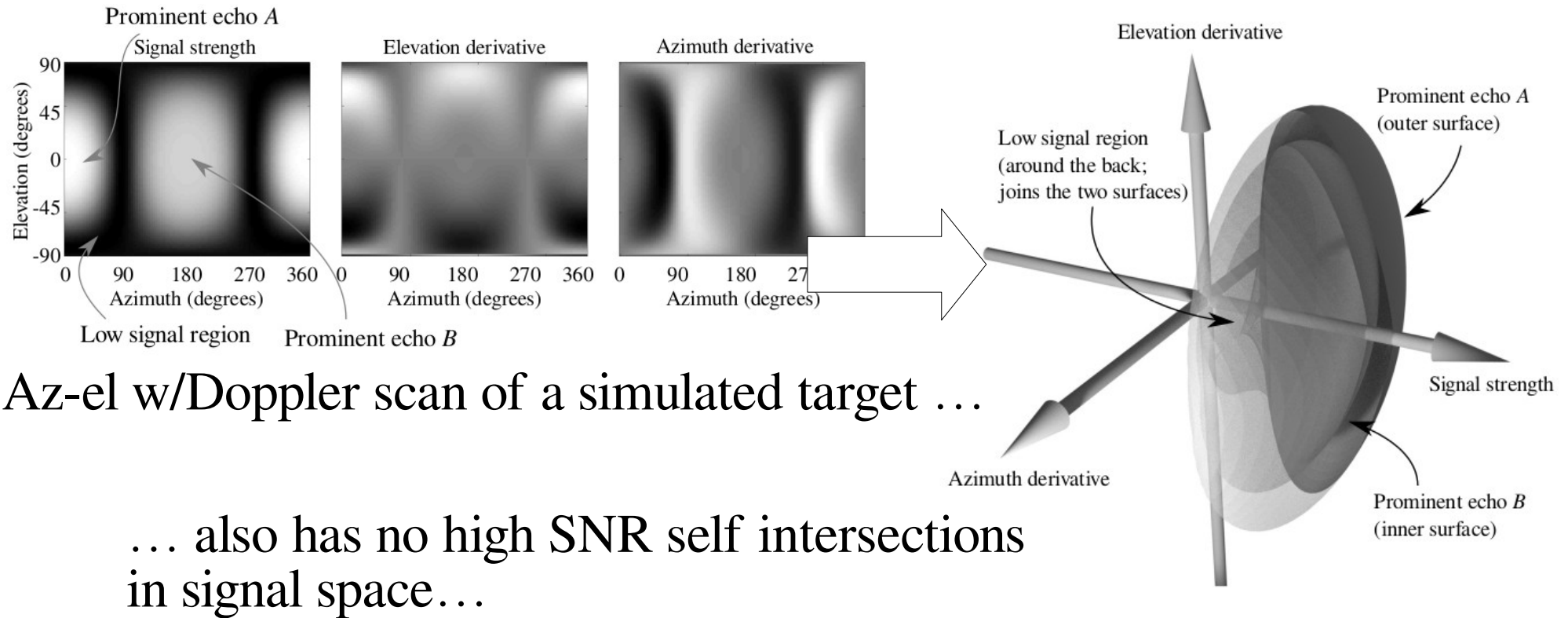
# Lab check: Styrofoam cup with lid



these loops don't really intersect

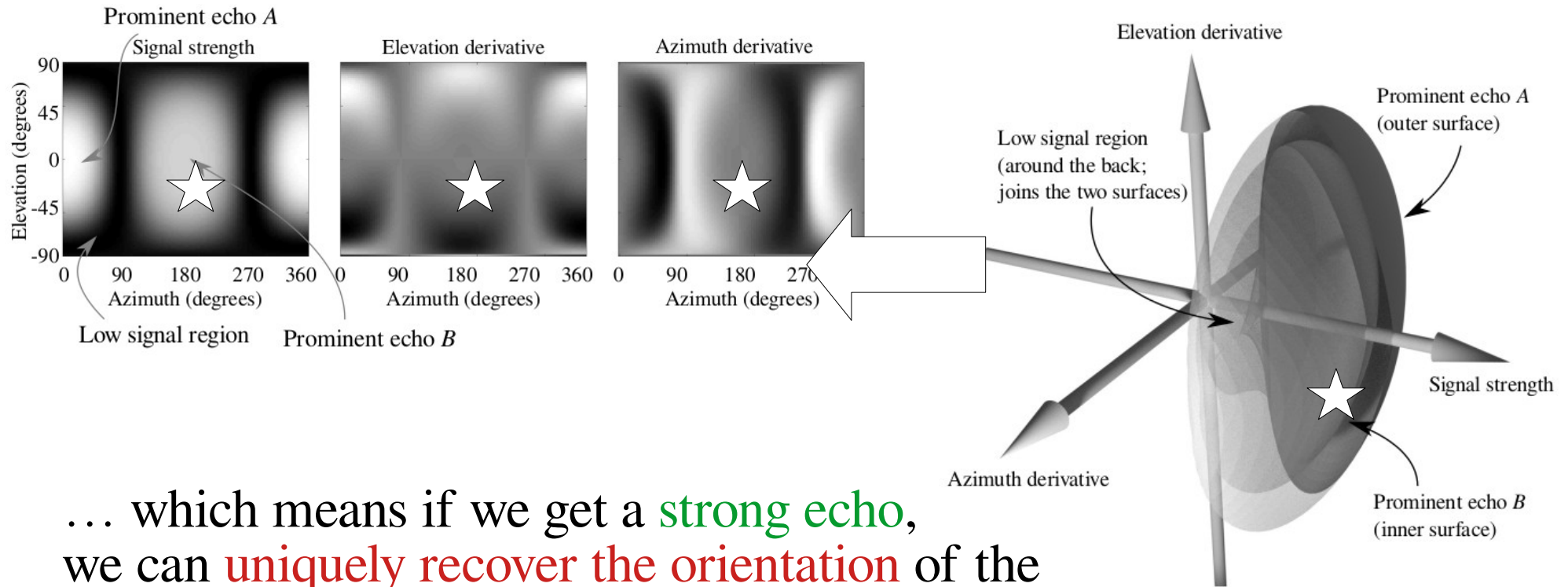
**Proposition 3.** *Let  $M$  be a smooth manifold and assume that  $n > (2 \dim M)$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the self-intersections in the signature space of  $v$  only occur at points where  $v(t) = 0$ .*

# An unexpected opportunity?



**Proposition 3.** *Let  $M$  be a smooth manifold and assume that  $n > (2 \dim M)$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the self-intersections in the signature space of  $v$  only occur at points where  $v(t) = 0$ .*

# An unexpected opportunity?

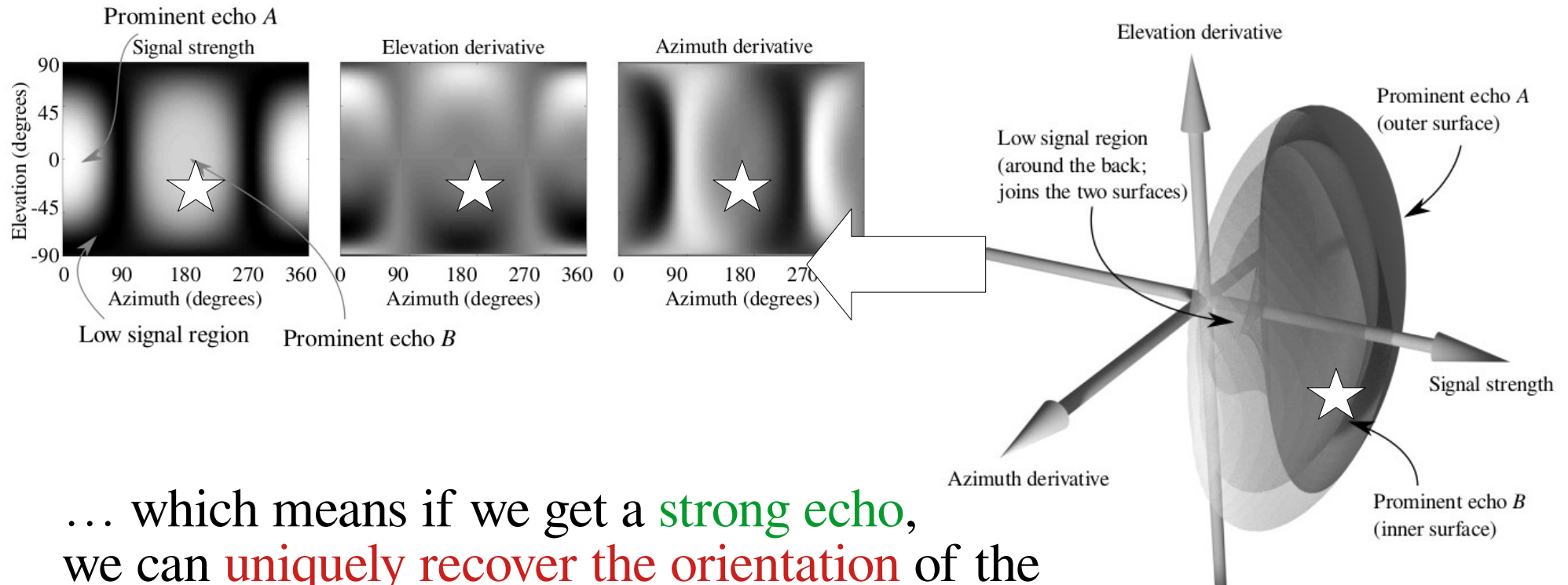


... which means if we get a **strong echo**,  
we can **uniquely recover the orientation** of the  
target **from one measurement!**

**Proposition 3.** *Let  $M$  be a smooth manifold and assume that  $n > (2 \dim M)$ . For a generic  $v$  in  $C_D^\infty(M, \mathbb{R}^n)$ , the self-intersections in the signature space of  $v$  only occur at points where  $v(t) = 0$ .*



# An unexpected opportunity?



... which means if we get a **strong echo**, we can **uniquely recover the orientation** of the target **from one measurement!**

This is a **topological version** of *monopulse estimation*, which appears to be completely novel (esp. the Doppler part)

Presently: Implementing this idea on data from collaborators!

# Next up

---

- Testing on real and realistic simulated data
  - Test bouquet-of-spheres hypothesis on collaborator-provided data
  - Test orientation-finding tools
- Tie back to topological structure theorems recently proven
  - Aim for geometric uncertainty quantification, especially around curvature, embedded reach, and the like...



# To learn more...

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<http://drmichaelrobinson.net>

Relevant papers:

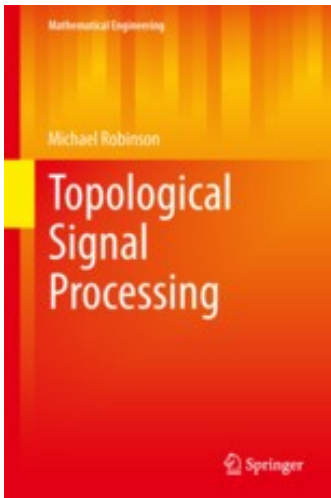
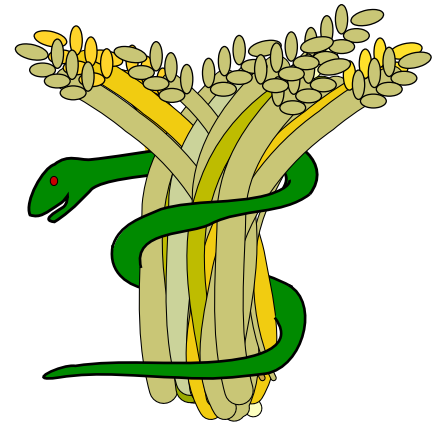
<https://doi.org/10.1121/10.0037085>

[arXiv:2205.11311](https://arxiv.org/abs/2205.11311)

<https://doi.org/10.1017/S0956792522000365>

Software:

<https://github.com/kb1dds>



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