

### Hybrid Concept Bottleneck Models

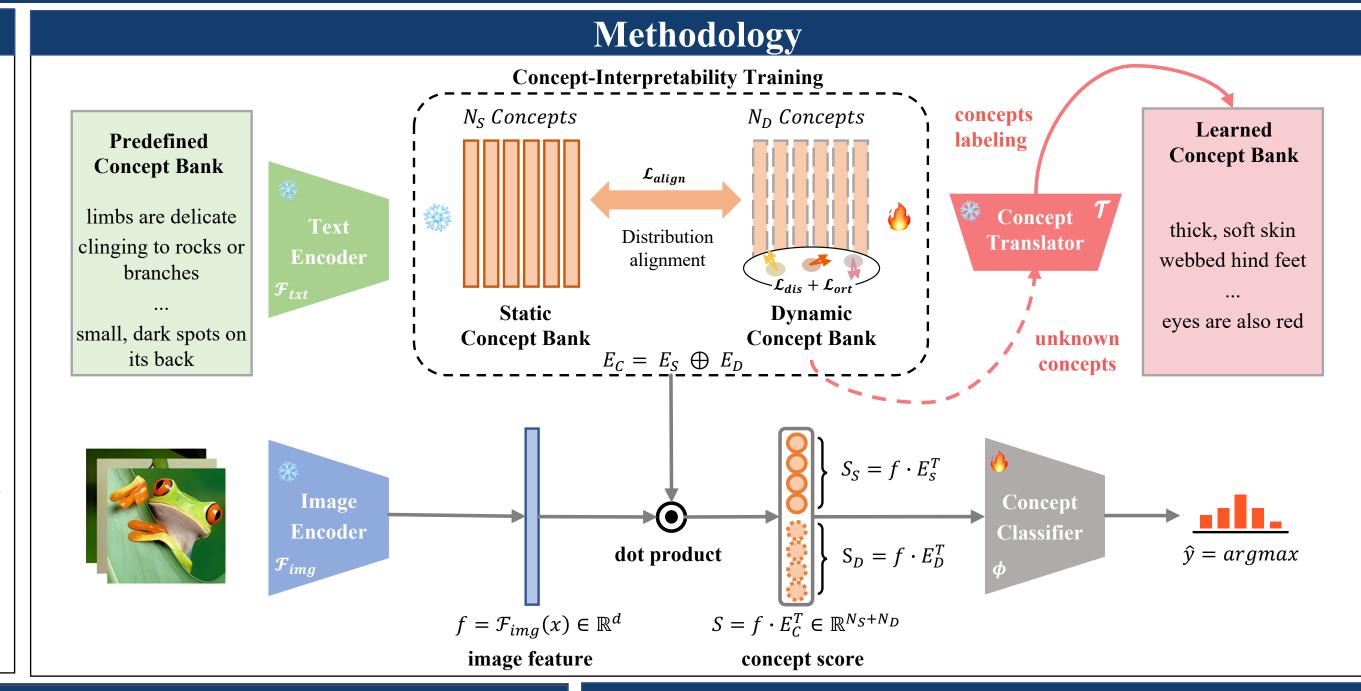
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# Input Image Predefined Concepts feeds on the ground eats insects, spiders yellow underparts Learned Concepts yellowish – throat high - pitched in the trees Lays 3-5 eggs

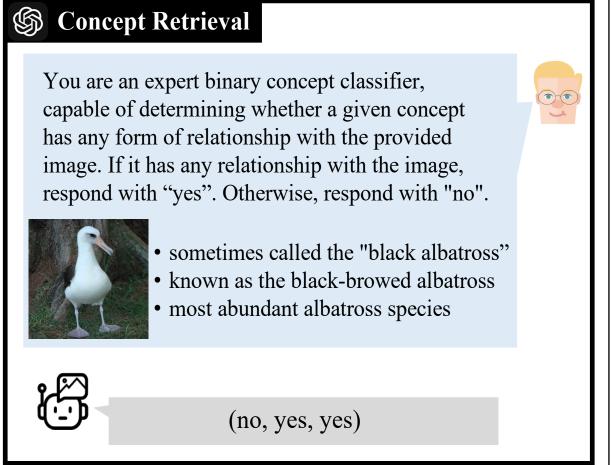
- The quality and completeness of the predefined concepts significantly affect the performance of CBMs.
- Innovative Hybrid Concept Bank: introduces a hybrid concept bank bank that combines static and dynamic concepts, allowing it to adapt and refine its interpretative capabilities dynamically.
- Integration of LLM Technologies:
  - 1) Using LLM for defining static concepts, translating new concepts
  - 2) Using VLM for evaluating the interpretability of concepts.



#### **Qualitative Analysis**

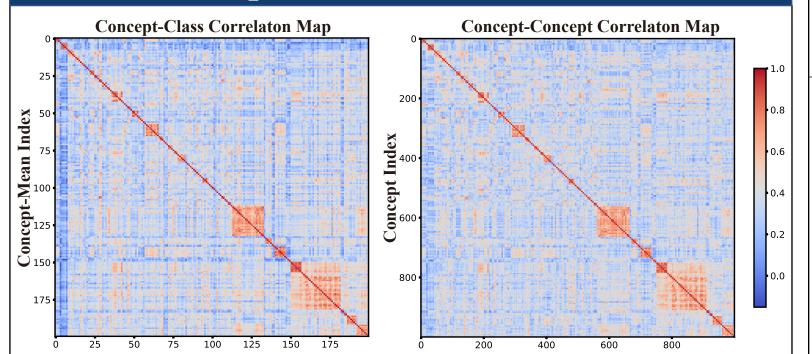
	Class Name	Sample	Case interpretability		
			static concepts	dynamic concepts	
CUB	Laysan Albatross		1.lays 4-6 white eggs 2.black and white striped body	1.black and white albatross 2.largest albatross species	
Food-101	Baklava		1.flaky, phyllo dough texture 2.filled with nuts and sweetened with syrup	1.light brown center 2.a food is arranged on a large platter	
Flower	Clematis		1.popular choice for making garlands and wreaths 2.borne on a climbing vine	1.large, a seed plant 2.violet	

The top-2 static and dynamic concepts for randomly selected classes across three datasets are presented, focusing on case interpretability.



The concept retrieval process with vision-language model GPT-40 is illustrated on the right.

#### **Concept Correlation Structures**

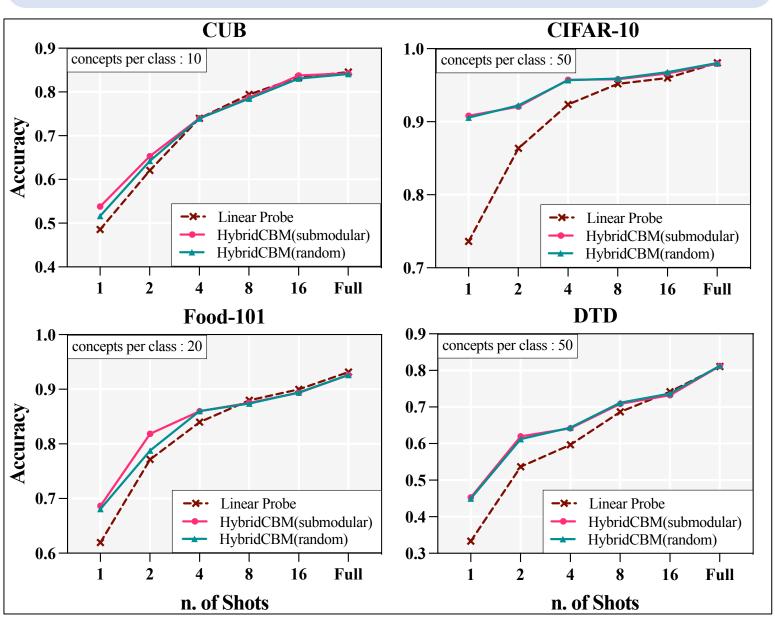


Comparison of dynamic concept correlation structures on the CUB dataset.

- Concept-Class concepts: our dynamic concepts align closely with their corresponding classes while effectively capturing diversity across classes.
- Concept-Concept concepts: highlights the diversity among concept with different classes.

#### **Quantitative Analysis**

Comparison of test accuracy between HybridCBM with submodular and random selection methods, and Linear Probe across 4 datasets.



## Evaluate the interpretability of HybridCBM's static and dynamic concepts through various metrics on two levels: Feature & Concept

		Feature Level		Concept Level	
Loss	Concept	Purity	Separation	Semantics	Precision@t
	Bank	(%)	(%)	(%)	(%)
	Static (submodular)	19.3	86.2	38.3	48.5
$\mathcal{L}_{cls}$	Static (random)	19.2	82.4	30.9	47.3
	Dynamic	0.1	100.0	0.01	0.05
$\mathcal{L}_{cls+dis}$	Dynamic	71.9	76.9	30.0	34.8
$\underset{+ort}{\mathcal{L}_{cls+dis}}$	Dynamic	72.0	77.1	28.9	34.1
$\mathcal{L}_{cls+dis}_{+ort+alig}$	Dynamic	39.8	80.0	32.8	46.2