

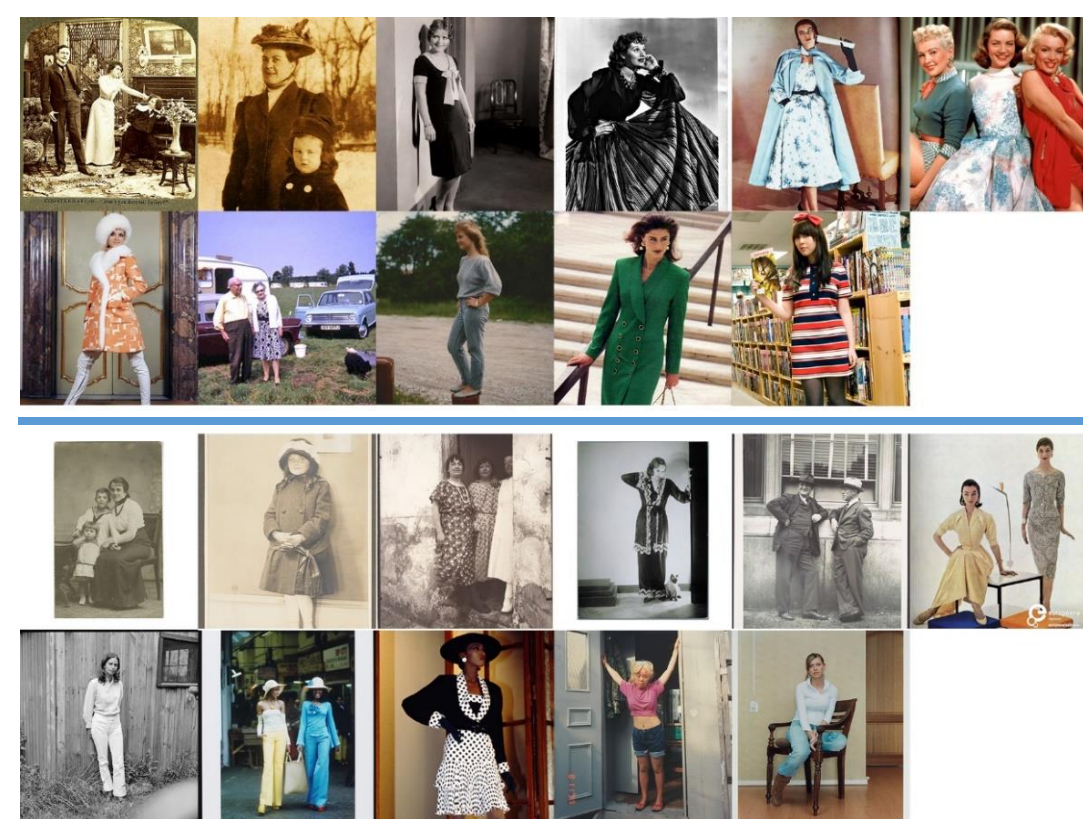
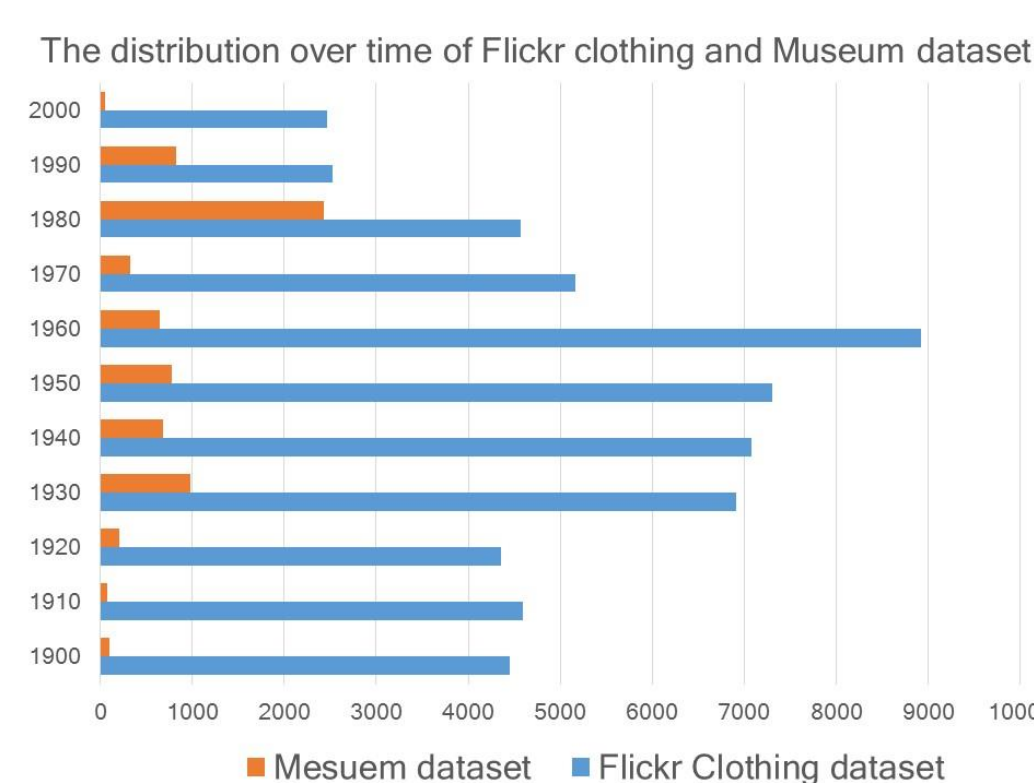
In this work, we explore deep learning methods for estimating when objects were made. Toward this goal, we both utilize features from existing deep networks and fine-tune new networks for temporal estimation and provide several analyses of what our networks have learned compare to the visual pattern discovery method.

## Contributions

- Deep learning approaches to temporal estimation
- Two datasets of dated photographs of clothing items.
- Analyses of what the fine-tuned networks for objects temporal estimation have learned, and comparison to the midlevel patterns learned by Lee et al.

## Dataset

- Flickr clothing dataset: 58350 images
- Museum clothing dataset: 9421 images



## Temporal estimation approaches

- Pre-trained network: Alexnet
- Fine-tuned network
  - fine-tuning and testing on CarDb

Methods	ACC (%)	MAE (years)	Methods	ACC (%)	MAE (years)
Lee et al.	48.96	8.56	Fc7 + SVM	50.11	7.77
Fc6 + SVM	53.78	7.24	Fc7 + SVR	50.16	8.10
Fc6 + SVR	51.57	7.80	Fine-tuned	65.04	4.91



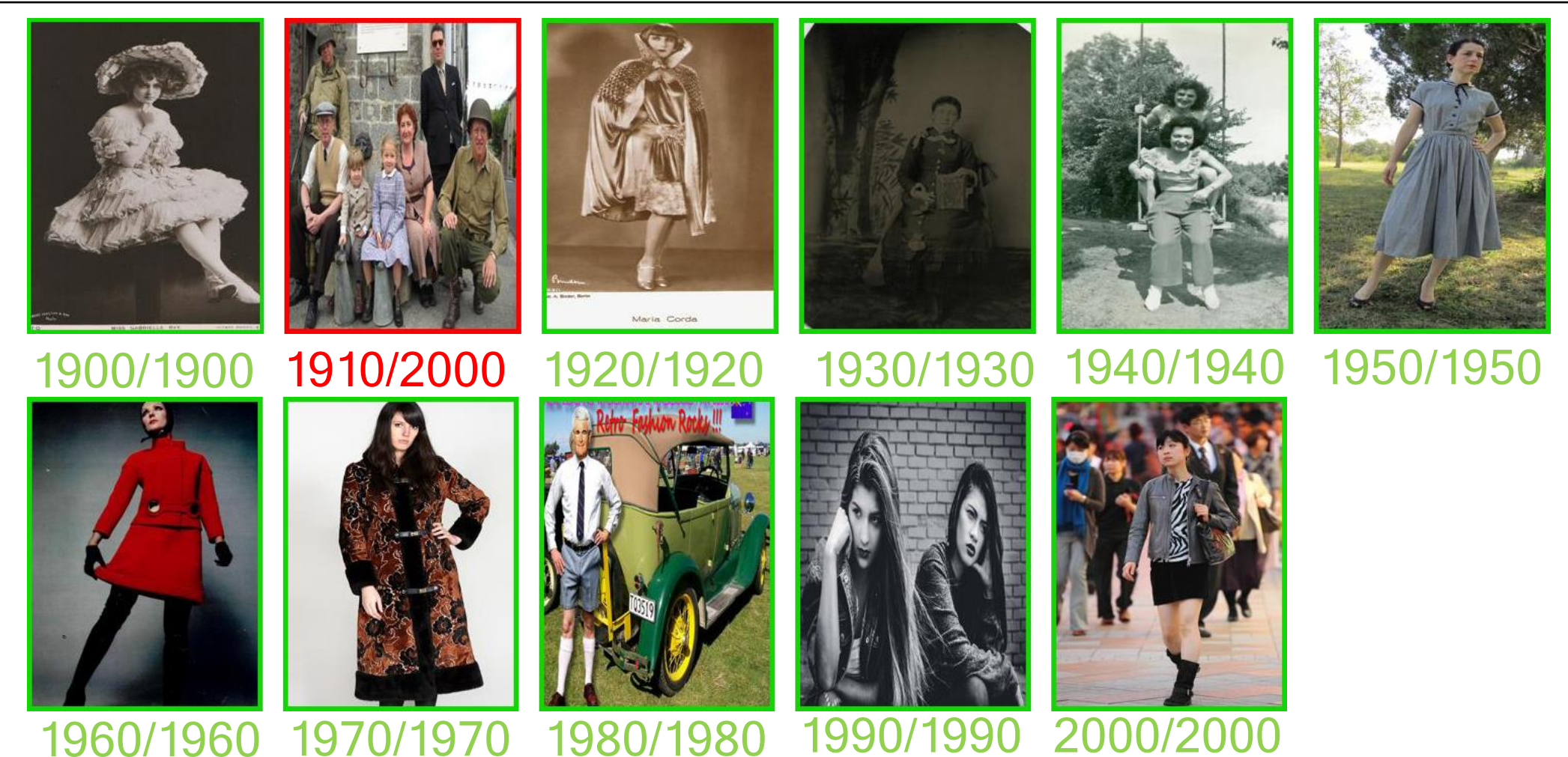
Lee et al.	1930	1950	1940	1930
Alexnet	1920	1920	1940	1950
Fine-tuned	1920	1920	1940	1950



Lee et al.	1920	1970	1960	1960
Alexnet	1950	1970	1960	1990
Fine-tuned	1960	1970	1980	1990

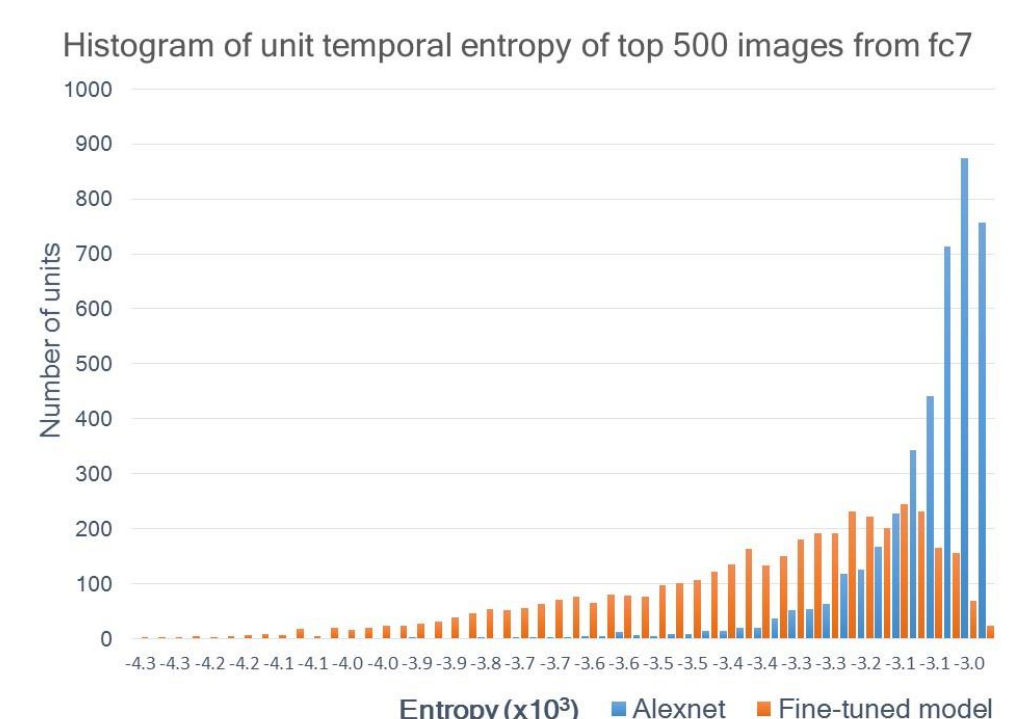
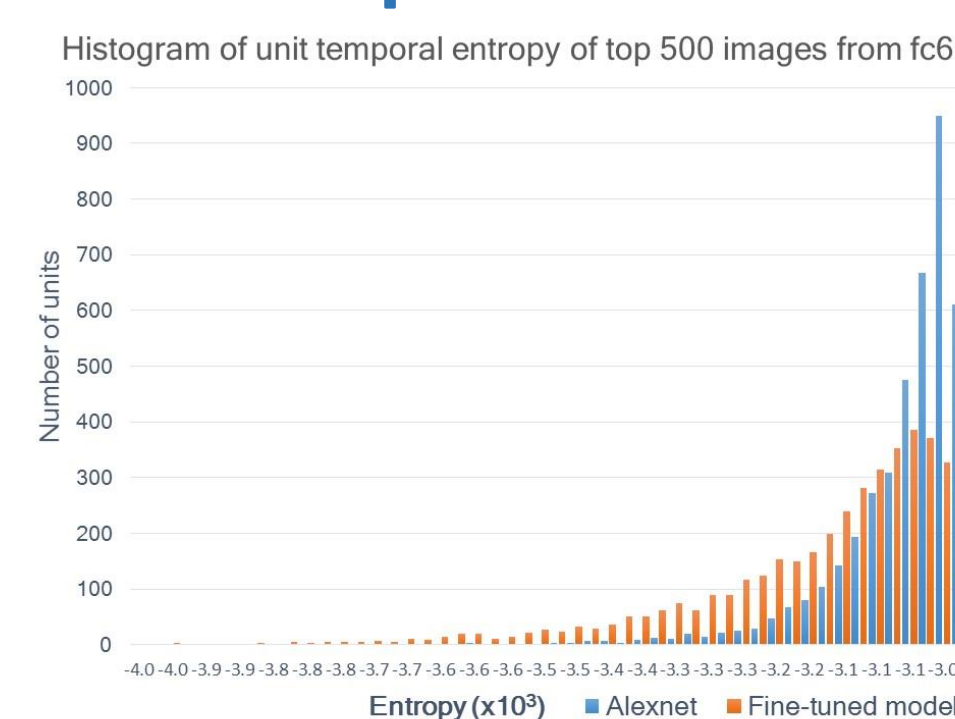
- fine-tuning on Flickr clothing dataset and testing on both Museum dataset and held out Flickr clothing dataset.
- fine-tuning 2 different models: color images and black/white images.

Methods	Museum dataset		Flickr clothing dataset	
	ACC (%)	MAE (years)	ACC (%)	MAE (years)
F.Palermo et al.	26.82	20.66	27.85	17.21
Fc6 + SVM	32.06	17.15	36.99	12.84
Fc6 + SVR	32.57	15.47	31.79	15.46
Fc7 + SVM	31.30	17.60	37.01	12.97
Fc7 + SVR	32.39	15.67	31.41	15.28
Fine-tuned(B/W)	43.11	13.89	37.23	13.12
Fine-tuned	45.89	11.01	42.57	11.10

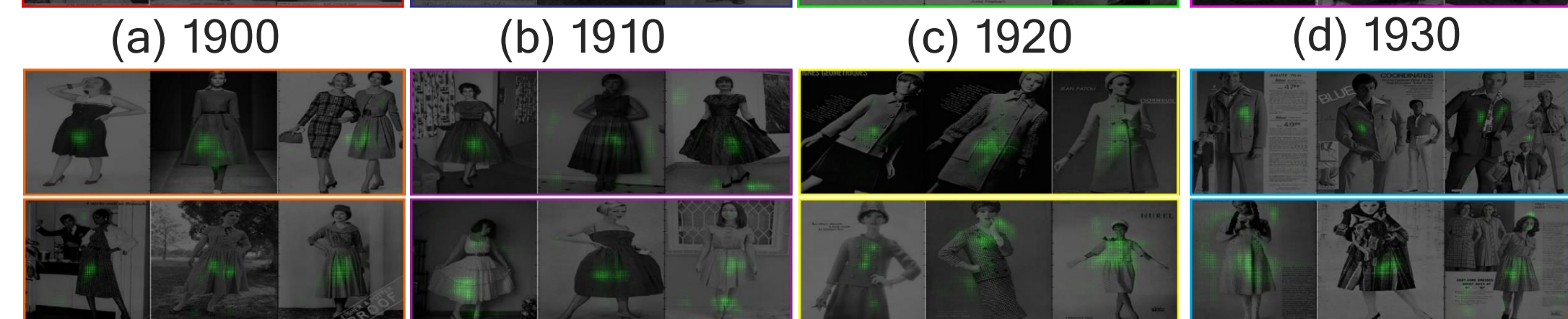
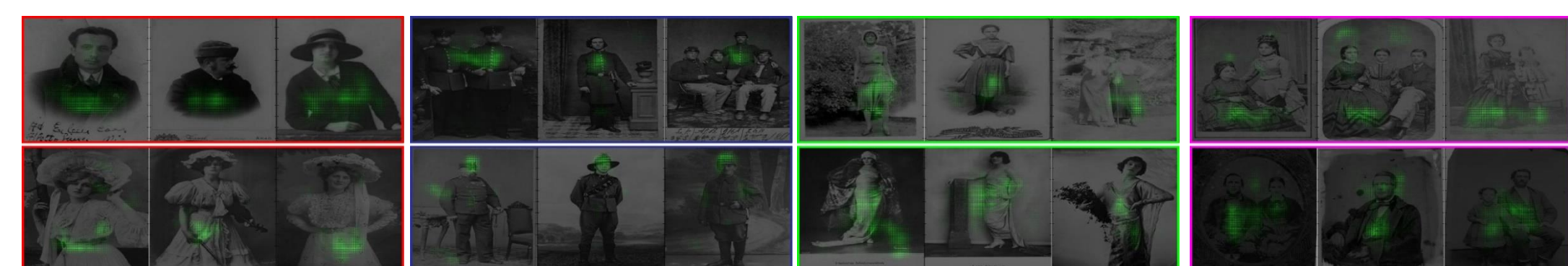
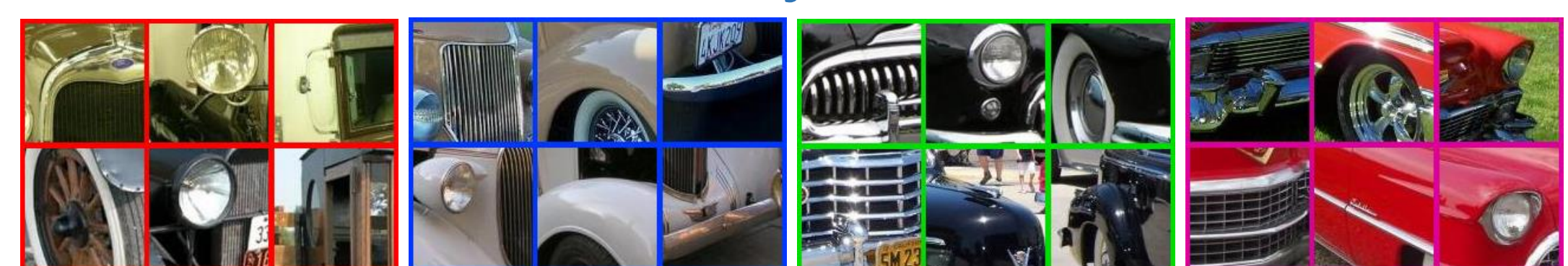


## Deep network analyses

### • Temporal - sensitive units



### • Unit activation analysis



### • Discriminative part correlation

