

# A Mishmash of Methods for Mitigating the Model Mismatch Mess

Andrew D. Ker<sup>a</sup> and Tomáš Pevný<sup>b</sup>

<sup>a</sup> Department of Computer Science, Oxford University, UK.

<sup>b</sup> Agent Technology Center, Czech Technical University in Prague, Czech Republic.

4th February 2014



## Image database to study model mismatch

- 9000 JPEG images from each of 9 Flickr users.
- All images share the same quantization table ( $qf = 85$ ).
- 8 different camera models from 5 manufacturers.
- Stego images embedded with nsF5 with payload 0.05 bpp.
- 7850 dimensional  $CF^*$  features used for steganalysis.

# The reality of model mismatch

		Testing actor								
		1	2	3	4	5	6	7	8	9
Training actor	1	0.00	0.01	0.13	0.23	0.01	0.10	0.05	0.04	0.08
	2	0.02	0.00	0.16	0.10	0.02	0.10	0.04	0.04	0.06
	3	0.36	0.33	0.02	0.38	0.24	0.23	0.36	0.29	0.31
	4	0.03	0.01	0.18	0.00	0.01	0.09	0.03	0.04	0.06
	5	0.04	0.01	0.11	0.07	0.00	0.09	0.03	0.02	0.06
	6	0.06	0.02	0.16	0.04	0.03	0.06	0.07	0.03	0.07
	7	0.01	0.00	0.21	0.10	0.00	0.11	0.01	0.02	0.06
	8	0.05	0.03	0.12	0.09	0.01	0.09	0.04	0.01	0.08
	9	0.03	0.01	0.19	0.17	0.01	0.11	0.08	0.04	0.05

Testing error of Fisher Linear Discriminant classifiers

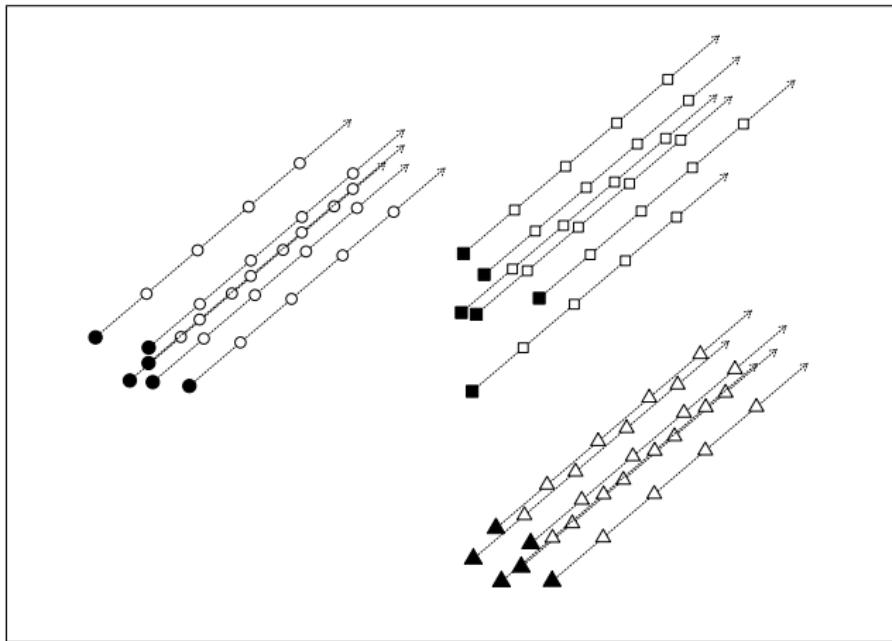
## How to measure error?

- Traditional error measures do not quantify the stability of the detector.
- Mean error rate  $\mu_1 = \frac{1}{k} \sum_i P_E^i$ .
- Root mean square error rate  $\mu_2 = \sqrt{\frac{1}{k} \sum_i (P_E^i)^2}$ .
- Maximum error rate  $\mu_\infty = \max_i P_E^i$ .

## Summary of the baseline

matched cases			mismatched cases		
$\mu_1$	$\mu_2$	$\mu_\infty$	$\mu_1$	$\mu_2$	$\mu_\infty$
0.0204	0.0315	0.0663	0.0981	0.1369	0.3887

## Mismatch due to shift



## Mean distances of covers

		Actor's whose centroid distance is measured...								
		1	2	3	4	5	6	7	8	9
...from other actor's centroid	1	0	0.59	3.93	12.58	0.66	1.90	2.82	0.72	0.79
	2	1.40	0	7.27	3.67	1.35	0.38	1.07	0.48	1.02
	3	12.84	12.46	0	19.02	10.56	9.25	13.92	12.00	12.58
	4	5.01	0.50	8.77	0	0.70	0.37	2.01	1.97	1.06
	5	0.49	1.09	2.54	5.39	0	2.17	2.29	2.56	2.00
	6	2.18	0.81	5.09	1.72	0.51	0	1.04	1.09	2.08
	7	0.26	0.35	7.53	2.92	0.53	0.76	0	0.49	0.10
	8	1.65	1.05	2.00	5.09	1.28	1.33	1.86	0	3.69
	9	0.26	0.61	4.82	3.31	0.44	0.95	1.26	1.30	0

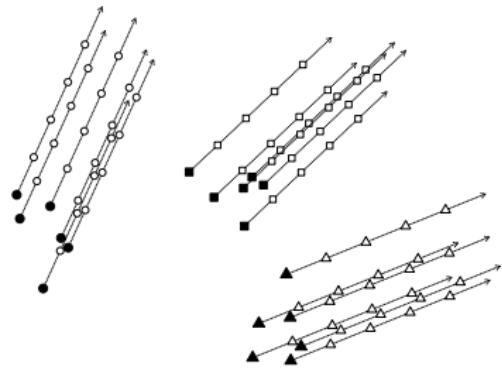
Correlation with errors:  $\rho = 0.91$     $\tau = 0.51$

## Centering the cover means helps

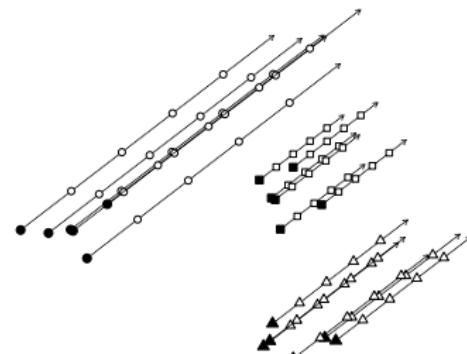
	matched cases			mismatched cases		
	$\mu_1$	$\mu_2$	$\mu_\infty$	$\mu_1$	$\mu_2$	$\mu_\infty$
Baseline	0.0204	0.0315	0.0663	0.0981	0.1369	0.3887
Subtracted mean				0.0691	0.0838	0.2332

# Mismatch due to angle and speed

images moves in different angles



images moves at different speed



$$\rho = -0.52, \tau = -0.44$$

$$\cos \alpha_{i,j} = \frac{w_i \cdot w_j}{\|w_i\| \|w_j\|}$$

$$\rho = -0.10, \tau = -0.09$$

$$r_{i,j} = \frac{\|w_i\| - \|w_j\|}{\sqrt{\|w_i\| \|w_j\|}}$$

## Calibrating for the different angle

- Re-embedding of a small payload and estimating the direction of the movement
- Training an ensemble of detectors, each on a different cover source
- weighting their vote based on the alignment measured by

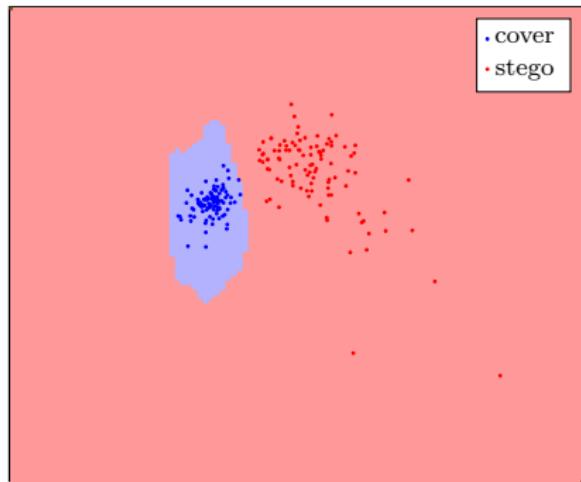
$$\blacktriangleright \text{angle } \cos \alpha_i = \frac{\mathbf{w}_i \cdot \boldsymbol{\delta}}{\|\mathbf{w}_i\| \|\boldsymbol{\delta}\|}$$

# Experimental results

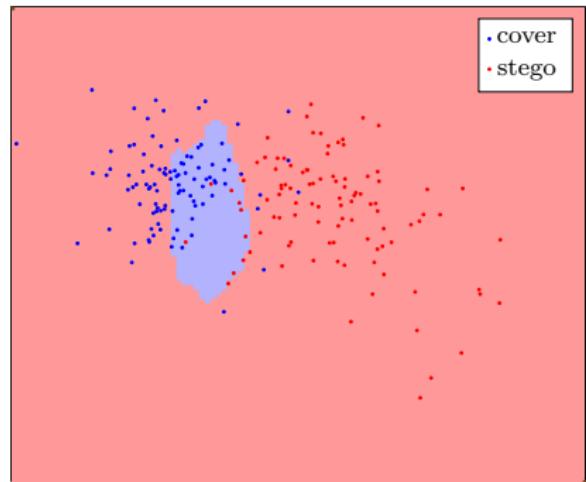
	matched cases			mismatched cases		
	$\mu_1$	$\mu_2$	$\mu_\infty$	$\mu_1$	$\mu_2$	$\mu_\infty$
Baseline	0.0204	0.0315	0.0663	0.0981	0.1369	0.3887
<i>Centered</i>				0.0691	0.0838	0.2332
<i>Ensemble voting</i>						
Equal weight				0.0439	0.0627	0.1366
Weight by $\cos \alpha_i$				0.0433	0.0593	0.1160
<i>Ensemble voting, centered</i>						
Equal weight				0.0391	0.0479	0.0776
Weight by $\cos \alpha_i$				0.0372	0.0468	0.0806

## Mismatch due to false certainty (1)

matched case



mismatched case



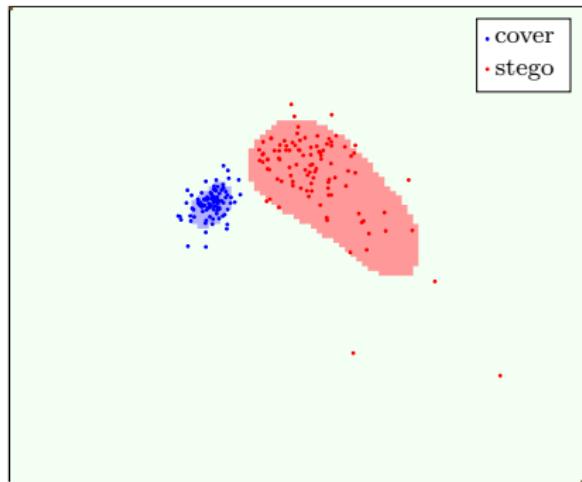
Decision areas of Binary Support Vector Machines

## Abstaining classifiers

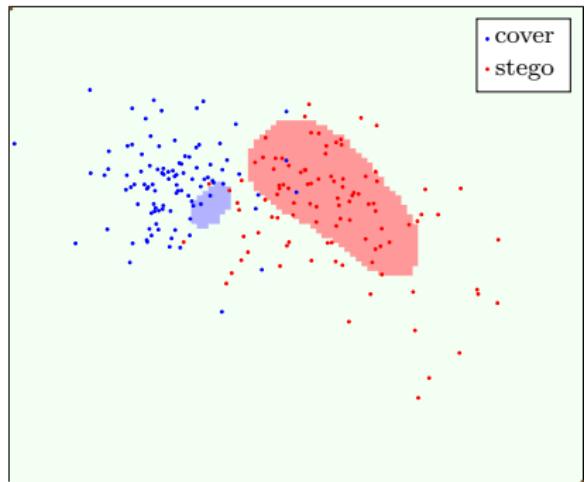
- “I don’t know” is
  - ▶ honest
  - ▶ potentially advantageous in pooled steganalysis.
- Implemented as a pair of 1-class Support Vector Machines (1-SVMs).
- 1-SVMs and 2-SVMs have hyper-parameters found by cross-validation.

## Mismatch due to false certainty (1)

matched case



mismatched case



Decision areas of abstaining SVMs

## Measuring error of abstaining classifiers

- The value of abstaining classifiers is in the pooled steganalysis.
- Introduce deflection coefficient

$$d = \frac{(1 - P_{\text{FP}} - P_{\text{FN}})\sqrt{1 - P_{\text{DK}}}}{\sqrt{P_{\text{FP}}(1 - P_{\text{FP}})} + \sqrt{P_{\text{FN}}(1 - P_{\text{FN}})}},$$

where  $P_{\text{DK}}$  is the probability of "don't know".

# Experimental results

mismatched cases				
	pooled error rates			$d_\infty$
	$P_{FP}$	$P_{FN}$	$P_{DK}$	
2-SVMs	0.165	0.059	0.000	0.465
1-SVMs	0.028	0.010	0.539	1.017

# Conclusion

- Uncorrected mismatched → error rates  $5\times$  matched case.
- Centering and matching direction → error rates  $1.5\text{-}2\times$  matched case.
- Possibility of answering “don’t know”.
- This is a mishmash of methods: no definitive answer.
- Research tends to be driven by the metrics:

are we chasing the right ones?