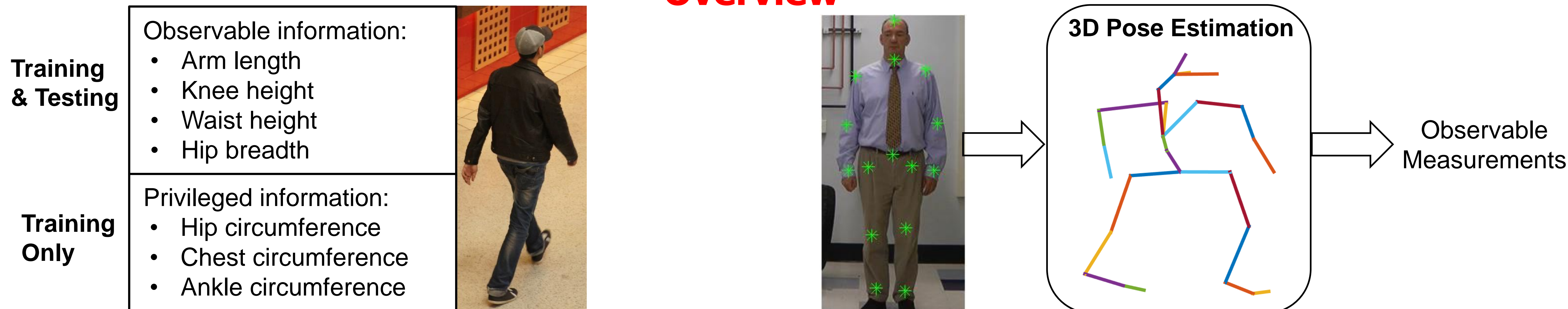


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## Overview



## Introduction

### Problem Statement

- Predict the gender using human metrology

### Motivation

- Explore the use of ratios of anthropometric measurements for gender estimation
- Exploit privileged information available during training

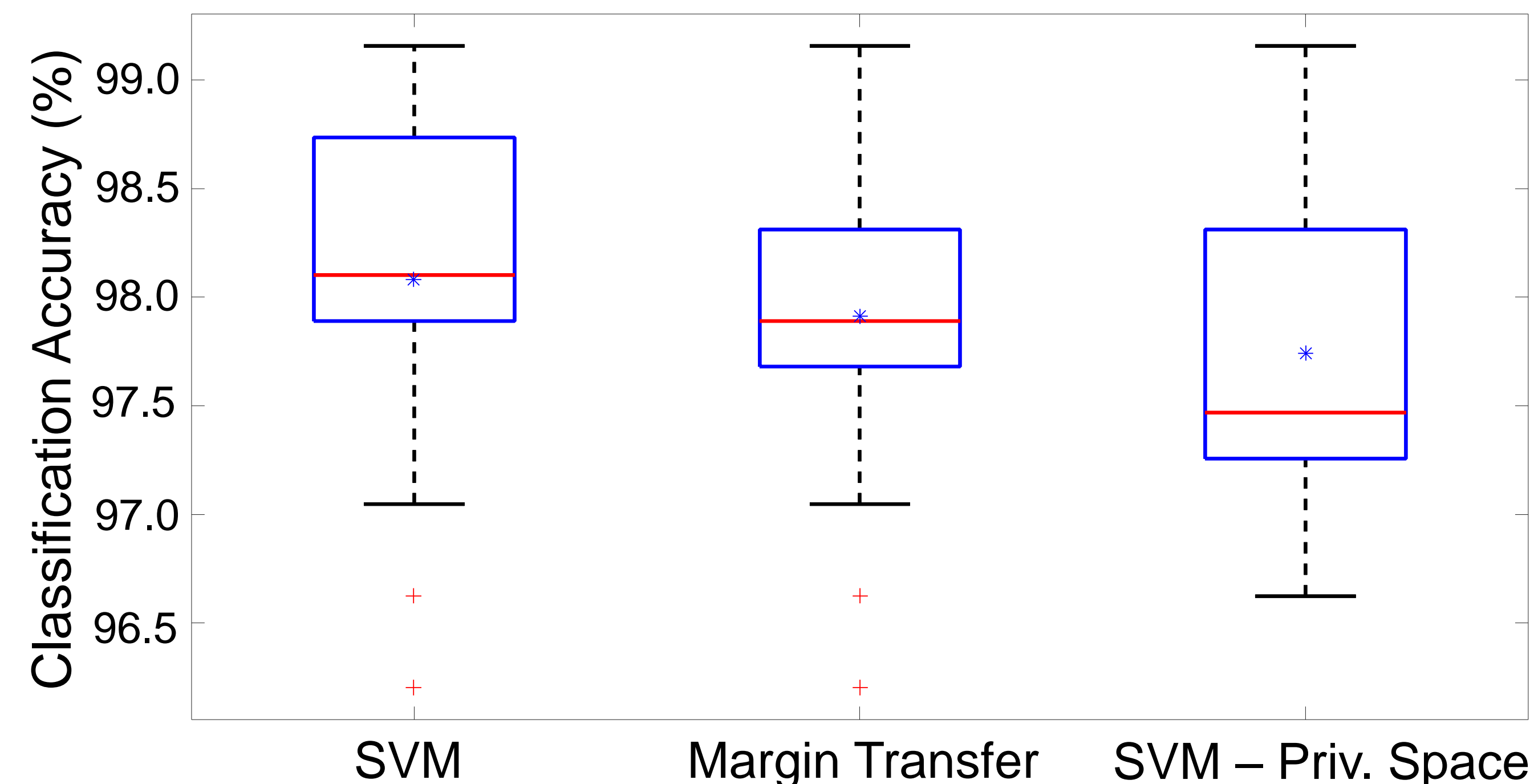
### Background

- Observable features: Information available at both training and testing
- Privileged features: Information available only at training time

## Results

### CAESAR dataset - Testing Features

	SVM	SVM+
Only observable (X)	97.61 ± 0.44	<b>98.18 ± 0.56</b>
Lower body observable (X <sub>L</sub> )	95.34 ± 0.74	<b>95.82 ± 0.81</b>
Upper body observable (X <sub>U</sub> )	<b>76.69 ± 2.98</b>	76.54 ± 2.95
Observable & Privileged (X+X*)	<b>99.10 ± 0.23</b>	-
Cao <i>et al.</i> [1]	99.37	-



### Set of features

Image Dataset	X	X <sub>L</sub>	X <sub>U</sub>
PaSC	71.37 ± 1.64	57.65 ± 2.82	58.06 ± 2.73
SARC3D	86.00 ± 2.00	78.00 ± 4.00	72.00 ± 4.00

## Contributions

- Using privileged information improves the classification accuracy
- Ratios of measurements are as discriminative as the actual values
- Predicting the gender from images using anthropometry is feasible

### References

[1] D. Cao, C. Chen, D. Adjeroh and A. Ross, "Predicting gender and weight from human metrology using a copula model," in Proc. 5<sup>th</sup> IEEE International Conference on Biometrics Theory, Applications and Systems, Washington, DC, USA, Sep. 23-26 2012, pp. 162-169.

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## Method

### Ratios of Anthropometric Measurements

- Original features:  $X = NxM_1, M_1 = \frac{F_1 x (F_1 - 1)}{2}$ ,  $F_1$  the number of original measurements
- Privileged features  $X^* = NxM_2, M_2 = \frac{F_2 x (F_2 - 1)}{2}$ ,  $F_2$  the number of privileged measurements
- Features are split to upper body ( $X_U$ ) and lower body ( $X_L$ ) sets

### Classification

$$\text{SVM: } \min_{\xi_1, \dots, \xi_N, w, b} \left\{ \frac{1}{2} \|w\|^2 + C \sum_{i=1}^N \xi_i \right\}$$

$$\text{s.t. } y_i (\langle w_i, x_i \rangle + b) \geq 1 - \xi_i, \xi_i \geq 0, i = 1, \dots, N$$

$$\text{SVM+: } \min_{\xi_1, \dots, \xi_N, w, b} \left\{ \frac{1}{2} (\|w\|^2 + \gamma \|w^*\|^2) + C \sum_{i=1}^N \xi_i (w^*, b^*) \right\}$$

$$\text{s.t. } y_i (\langle w_i, x_i \rangle + b) \geq 1 - \xi_i (w^*, b^*) \quad \xi_i (w^*, b^*) \geq 0, \\ i = 1, \dots, N$$

$$\text{Margin Transfer: } \min_{\xi_1, \dots, \xi_N, w, b} \left\{ \frac{1}{2} \|w\|^2 + C \sum_{i=1}^N \xi_i \right\}$$

$$\text{s.t. } y_i (\langle w_i, x_i \rangle + b) \geq \rho_i - \xi_i, \xi_i \geq 0, i = 1, \dots, N$$

w the weight vector, b the bias parameter,  $\xi_i$  is the slack, and C denotes the penalty parameter