

Known-Component Registration

Thomas Yi

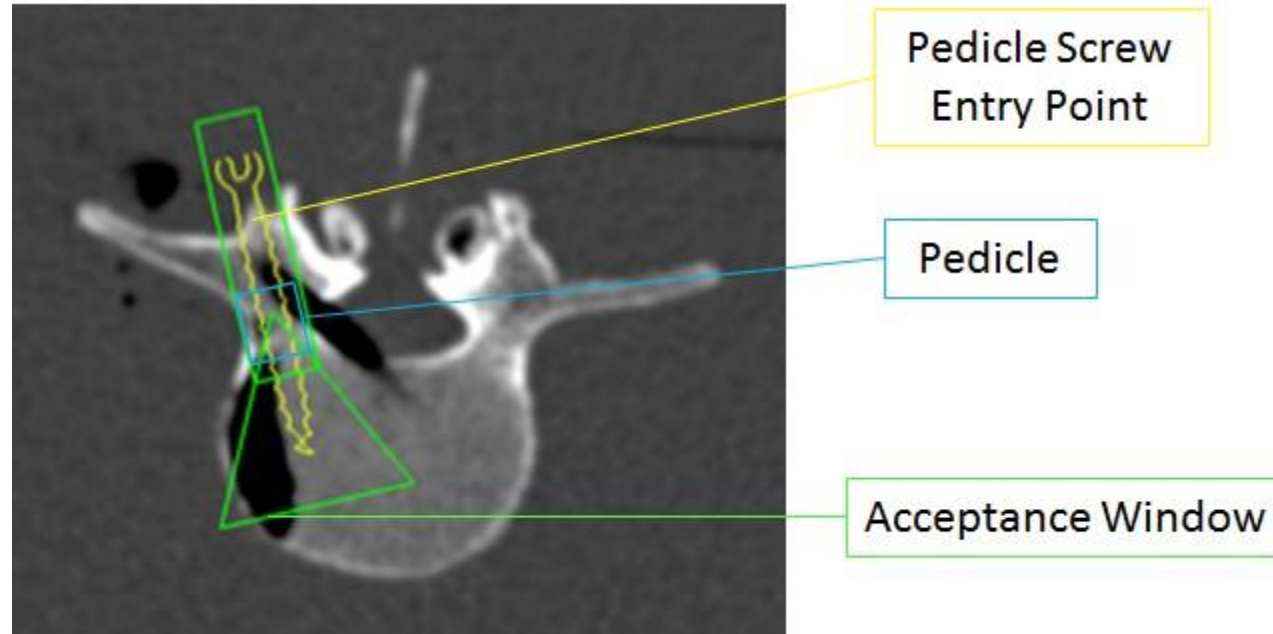
Group 1: High-Precision Drill Guide Placement with the UR5

Mentors: Ali Uneri, Jeff Siewerdsen



Project Overview

- Goal: to improve pedicle screw placement procedures
- Current standard of care involves “free-hand” placement of screw
- Optical Tracker Solution vs. 3D-2D Registration Solution



Paper Selection

“Known-component 3D-2D registration for quality assurance of spine surgery pedicle screw placement”

- Describes namesake technique: known-component registration (KC-Reg)
- Allows for localization of a “known component” in some 3D space (ie. preoperative CT)



Background

- Known Component
 - Surgical tools (fixation hardware, guide wires, needles, screws) that are structurally known beforehand
 - Varying degrees of structural knowledge
- 3D-2D Registration
 - Iteratively match intraoperative 2D radiographs to digitally reconstructed radiographs (DRRS) from a preoperative 3D CT volume.
 - Maximize image similarity



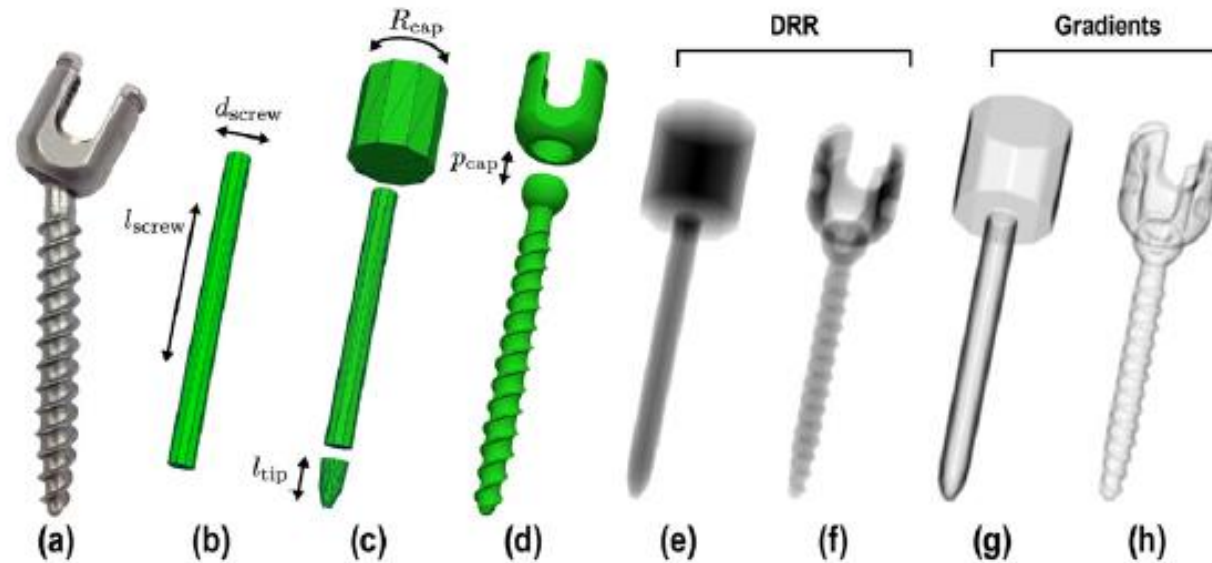
Significance

- Could remove the necessity for an optical tracking in our current system
- Removes physical clutter from operating space
- Eliminates problems associated with tracking reliability
- Tool simplification



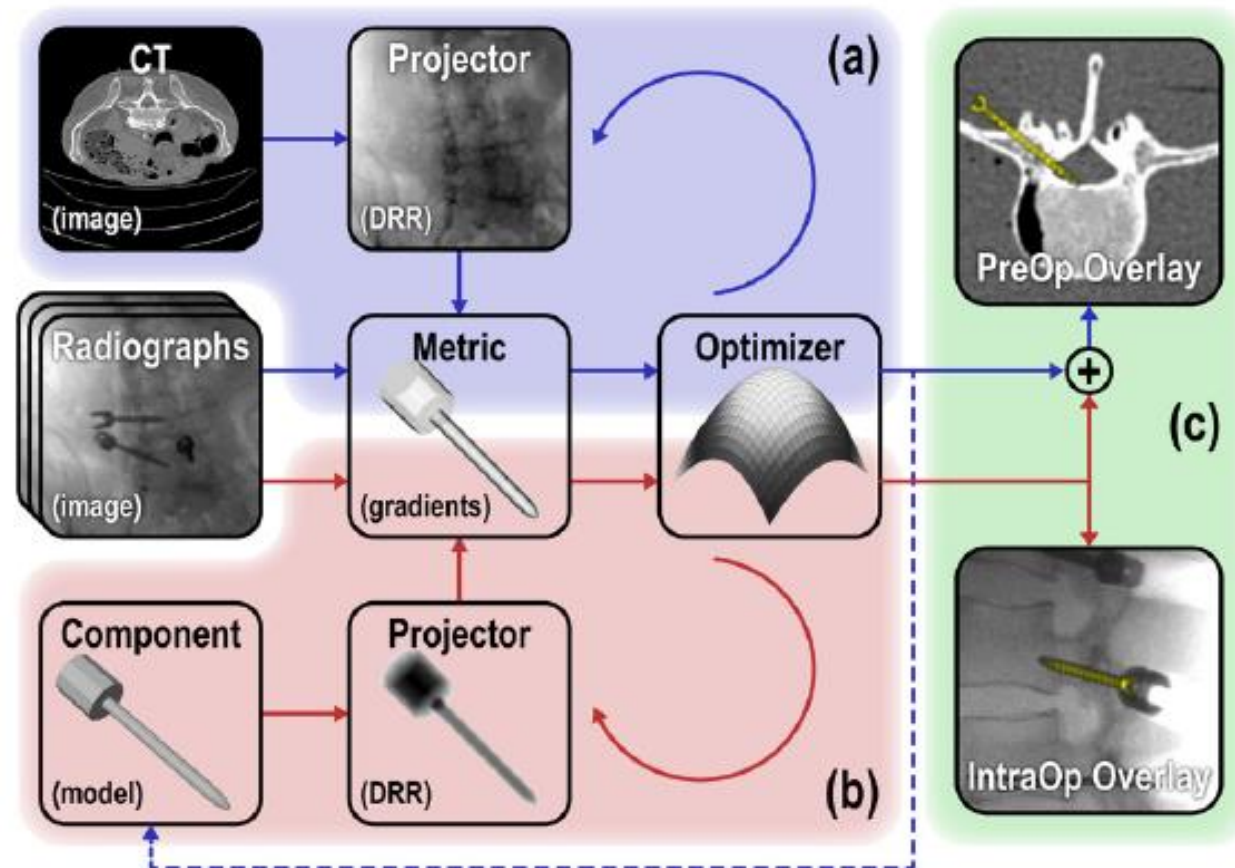
Background: Known Components

- Parametrically Known Component (pKC)
 - pKC1 – most simple parametrization
 - pKC2 – more complex, multi-component parametrization
- Exactly Known Component (eKC)
 - CAD model



Flow of Technique

- KC-Reg goes along with standard 3D-2D registration



Mathematical Methods

- Maximizing the similarity metric

Gradient Correlation (Penney 1998)

Directional Gradients of Fixed Radiograph (f) and Moving DRR (m)

$$GC(f, m) = \frac{1}{2}(CC(\nabla_x f, \nabla_x m) + CC(\nabla_y f, \nabla_y m))$$

Average Normalized Cross Correlation

(i, j) 2D Image Pixel Coordinates

$$CC(f, m) = \frac{\sum_{i,j} (f_{ij} - \bar{f})(m_{ij} - \bar{m})}{\sqrt{\sum_{i,j} (f_{ij} - \bar{f})^2} \sqrt{\sum_{i,j} (m_{ij} - \bar{m})^2}}$$

Maximize Total GC
Across All Projection
Views

$$\hat{\Phi} = \arg \max_{\Phi} \sum_{\theta} GC(f_{\theta}, m_{\theta}(\Phi))$$

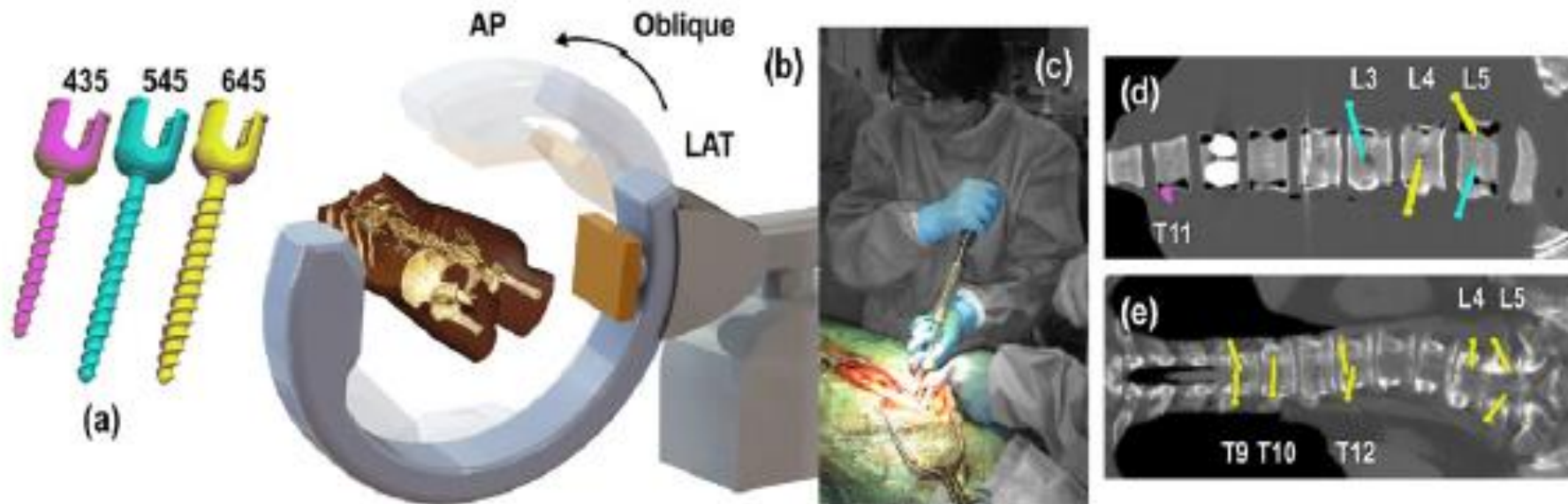
Moving DRR Computation

$$m_{\theta}(\Phi) = \mathcal{P}_{\theta} T(\Phi) \circ M$$



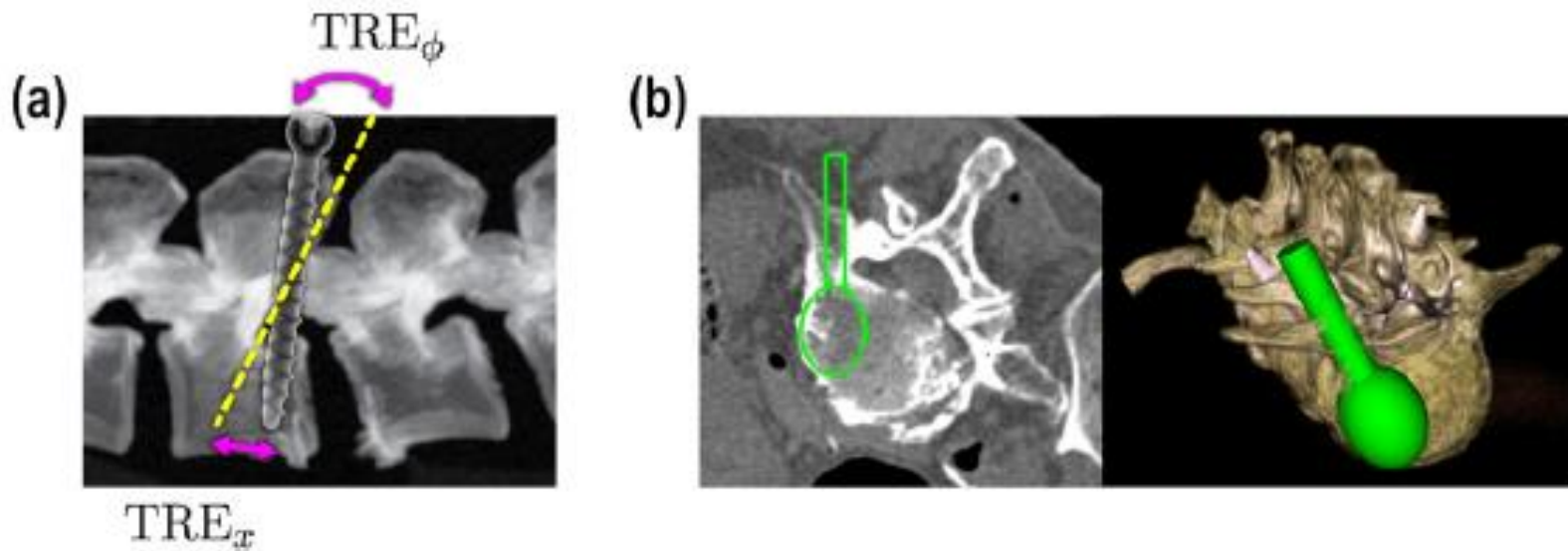
Experimental Setups

- Anthropomorphic torso phantom with 5 pedicle screws
- Human torso cadaver with 8 pedicle screws
- Intraoperative radiographs obtained with mobile C-arm as shown
- QA analysis: geometric accuracy, device verification, visualization relative to acceptance window



Geometric Accuracy of Registration

- TRE computed in terms of translational and rotational components
 - Ground truth determined from 3D-2D registrations using all available views (200 projections acquired over a semicircular arc using mobile C-arm)



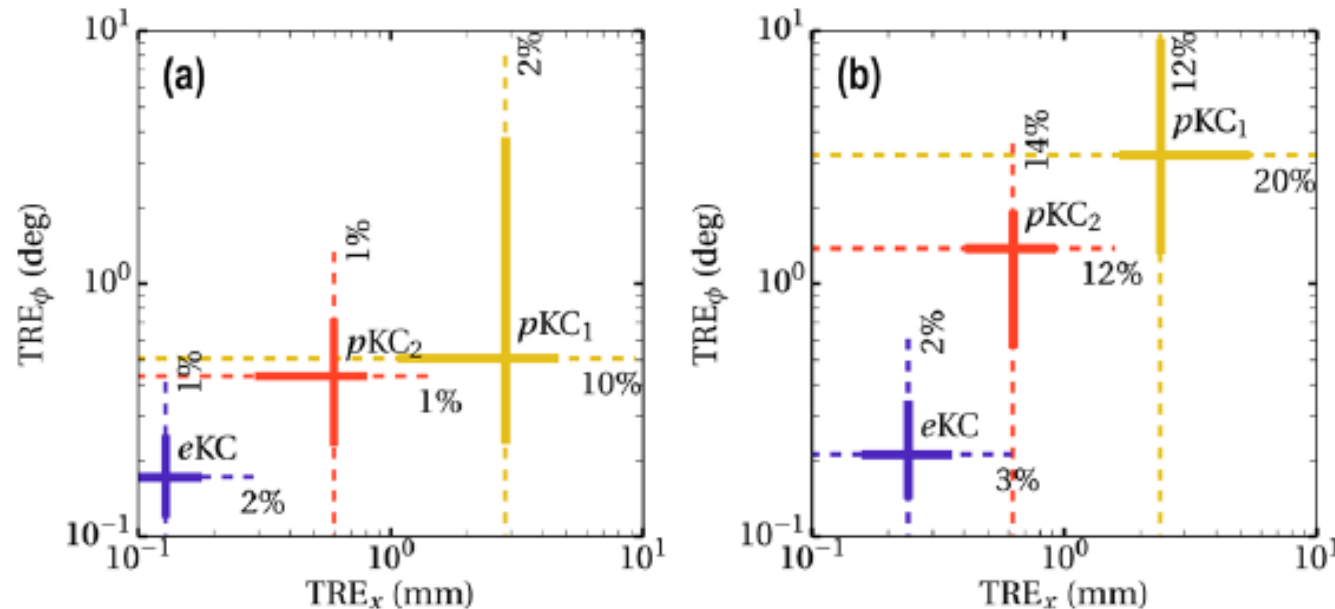
(\hat{R}, \hat{t}) registration estimate

$$TRE_x = \hat{t} - t_2$$

$$TRE_\phi = \cos^{-1} \frac{\hat{R}_z \cdot R_z}{|\hat{R}_z| \cdot |R_z|}$$

Geometric Accuracy Results

- Higher-order known components offer lower TRE
 - 92% of 40 repeat registrations per target screw within gold standard TRE accuracy levels of <1 mm in translation and <5 degrees in rotation for eKC
 - Median translational and rotational errors for eKC were 0.2 mm and 0.2 degrees
- Cadaver presents higher TRE likely due to deforming soft tissue and more complex gradients in real anatomy



Device Verification & Results

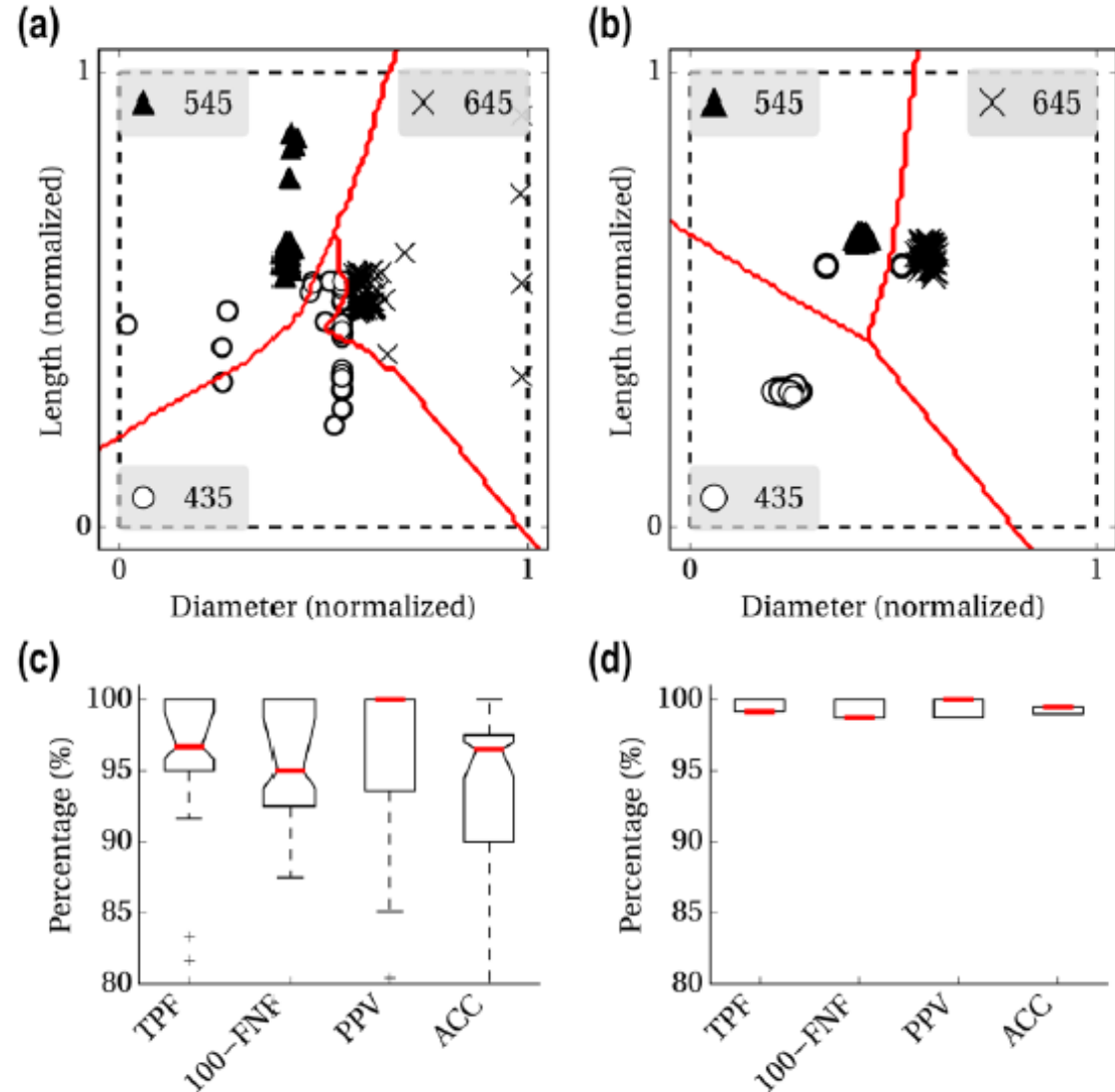
- Extension of KC-Reg methodology to detect instances in which device in 2D intraoperative image differs from that specified in planning
- 200 data samples of registration solution output parameters (length, diameter) used to train multi-class learning-based classifier
- Decision boundaries for classification reasonable in both cases
- Higher-order pKC2 had better classification results (99.3% vs pKC1 92.9%)

$$\text{TPF} = \text{TP} / \text{P}$$

$$\text{FNF} = \text{FN} / \text{P}$$

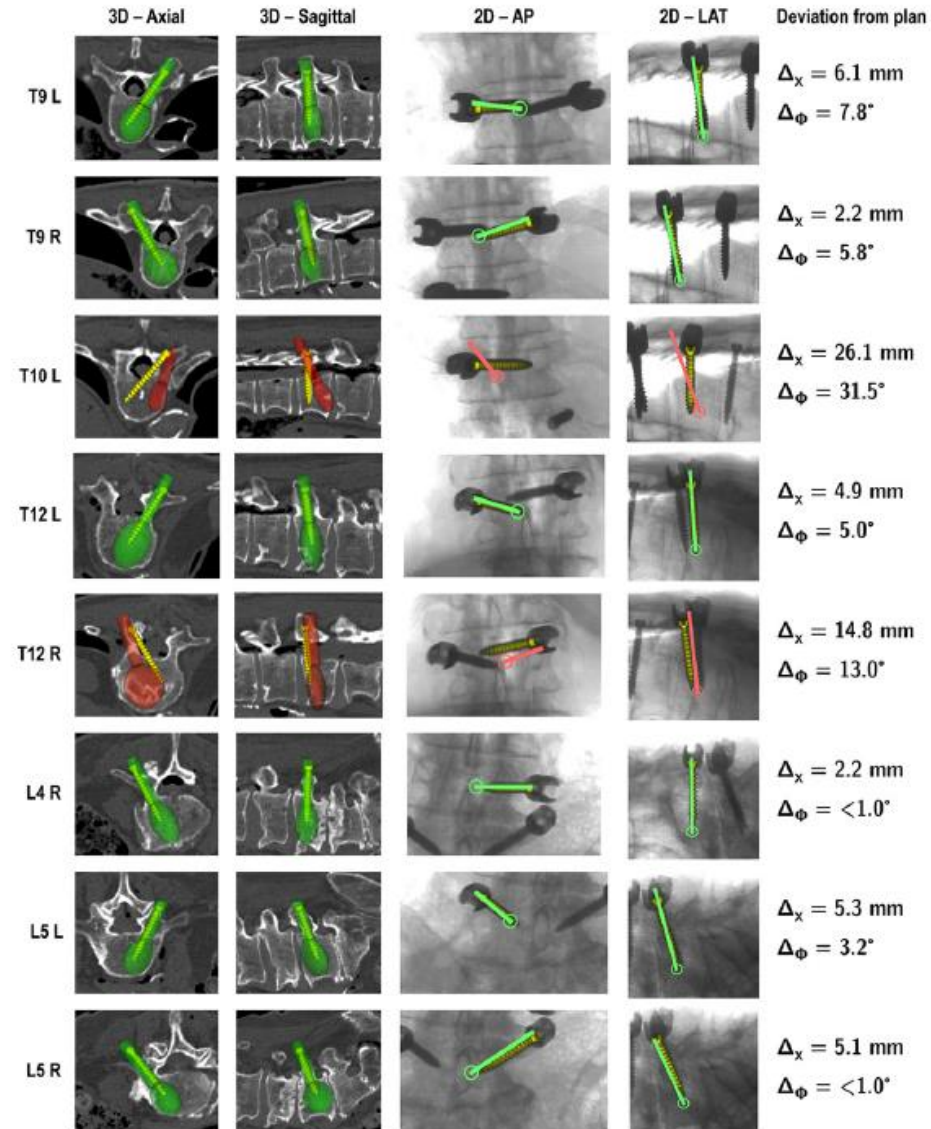
$$\text{PPV} = \text{TP} / (\text{TP} + \text{FP})$$

$$\text{ACC} = (\text{TP} + \text{TN}) / (\text{P} + \text{N})$$



Visualization Within Acceptance Window

- Screws purposely misplaced
- Assessment if screws were within acceptance window
- Acceptance window defined around planned trajectory
- KC-Reg result was able to correctly identify whether or not screws were properly placed



Assessment

Pros

- Overall well-described methodology surrounding the experiments
- Detailed in outlining how models were parametrized
- Powerful potential in substituting out optical tracking for described techniques
- Results very cleanly and intuitively discussed

Cons

- Organization slightly misleading
 - Could have been improved with more initial description on known-components
- Perhaps could have presented some additional base information surrounding the classification methods



Questions?

