



JOHNS HOPKINS
MALONE CENTER *for*
ENGINEERING *in* HEALTHCARE

Machines Teaching People

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How is skill acquired?

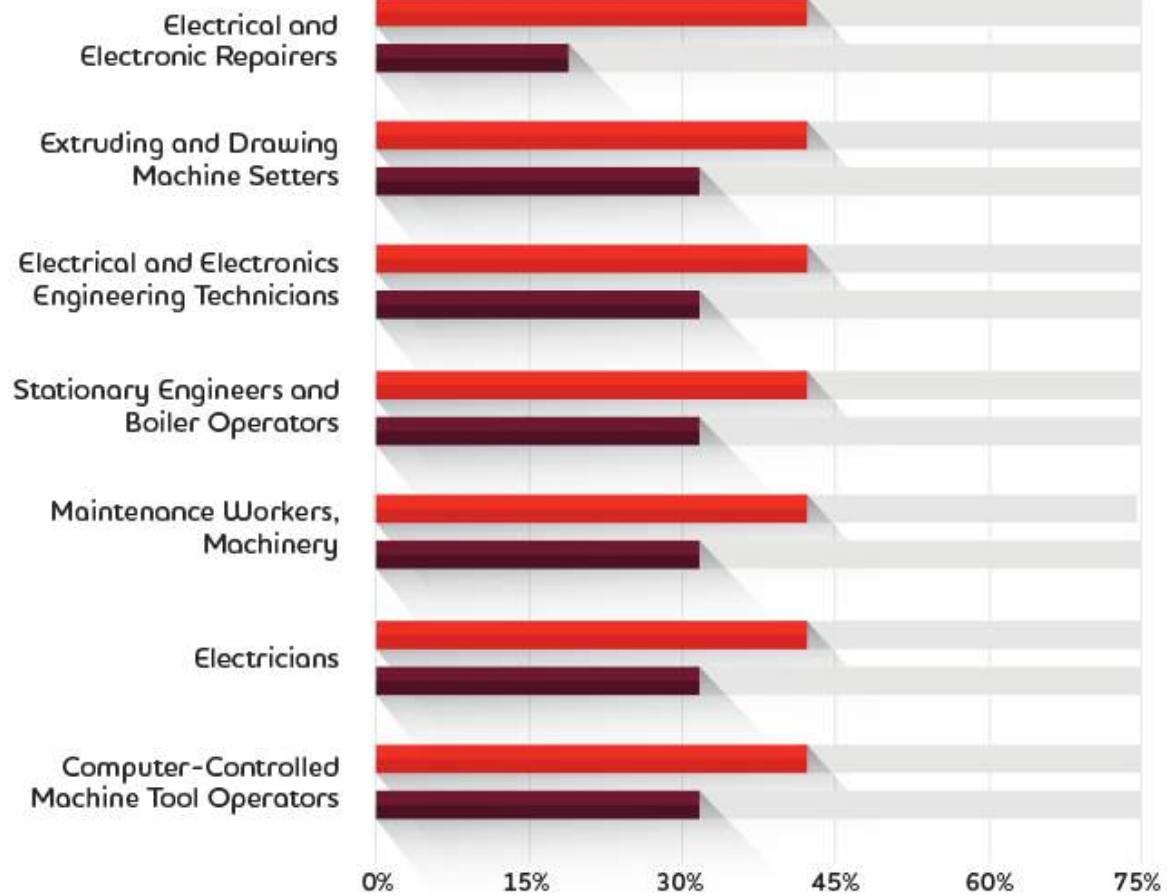
Gladwell, Malcolm. *Outliers: The story of success*. Penguin UK, 2008.

Campitelli, Guillermo, and Fernand Gobet. "Deliberate Practice Necessary But Not Sufficient." *Current Directions in Psychological Science* 20, no. 5 (2011): 280-285.

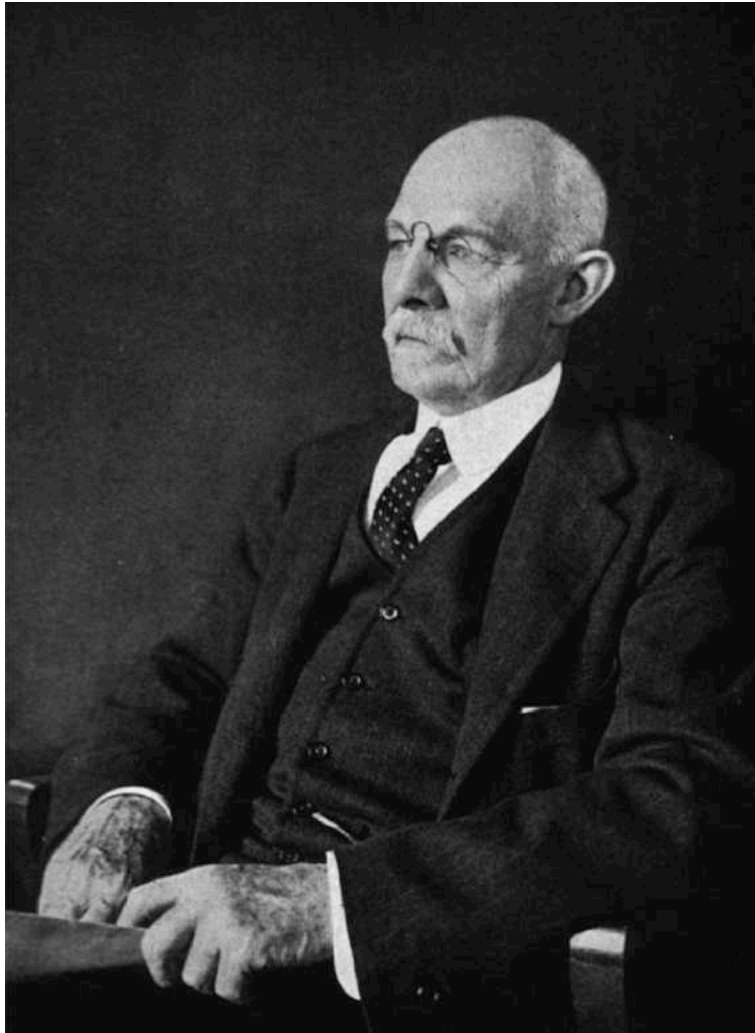


The Oldest Skilled Trade Jobs

■ Share of 45+ Workers ■ Share of 55+ Workers



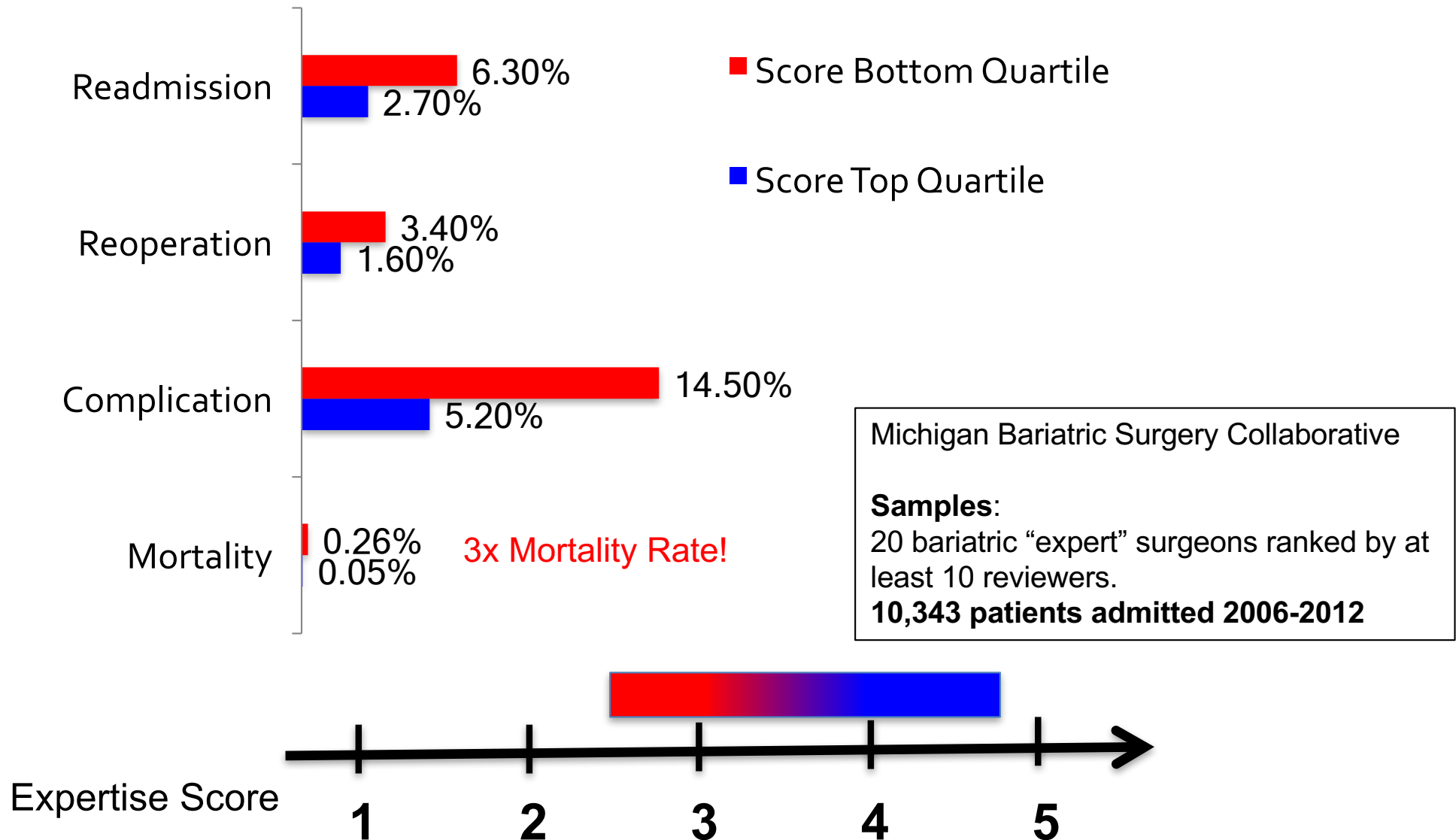
Skill Acquisition: Apprenticeship



William S. Halsted, JHU 1889
“See one, do one, teach one.”

C. Franzese and S. Stringer. The Evolution of Surgical Training: Perspectives on Educational Models from the Past to the Future. *Otolaryngologic Clinics of North America*, 40(6):1227–1235, 2007.

Training and Skill Impacts Patient Outcomes





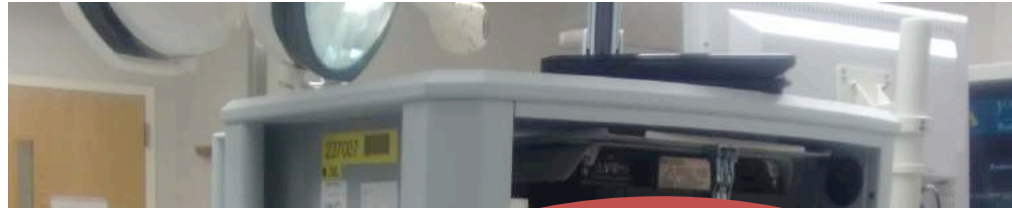
Credit: Intuitive Surgical



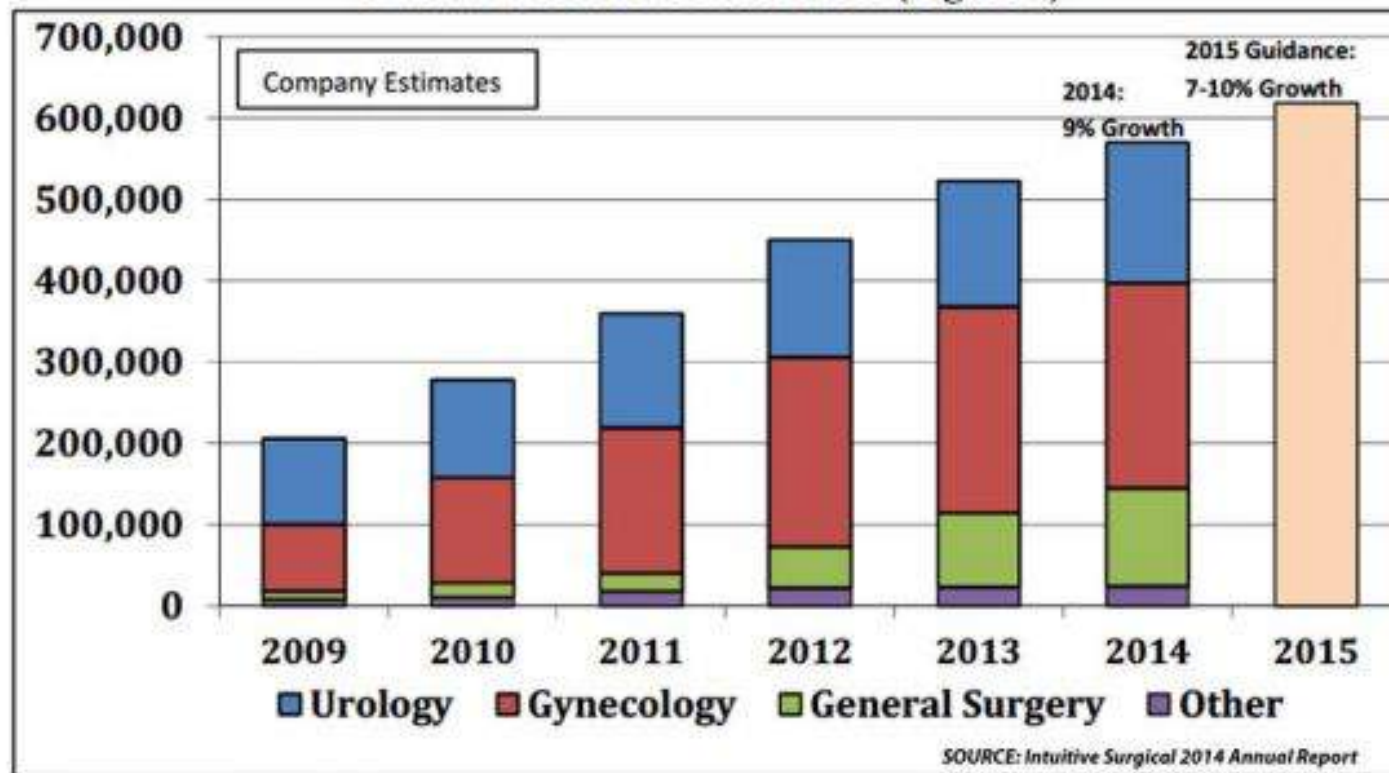
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CIRL Computational Interaction
and Robotics Laboratory

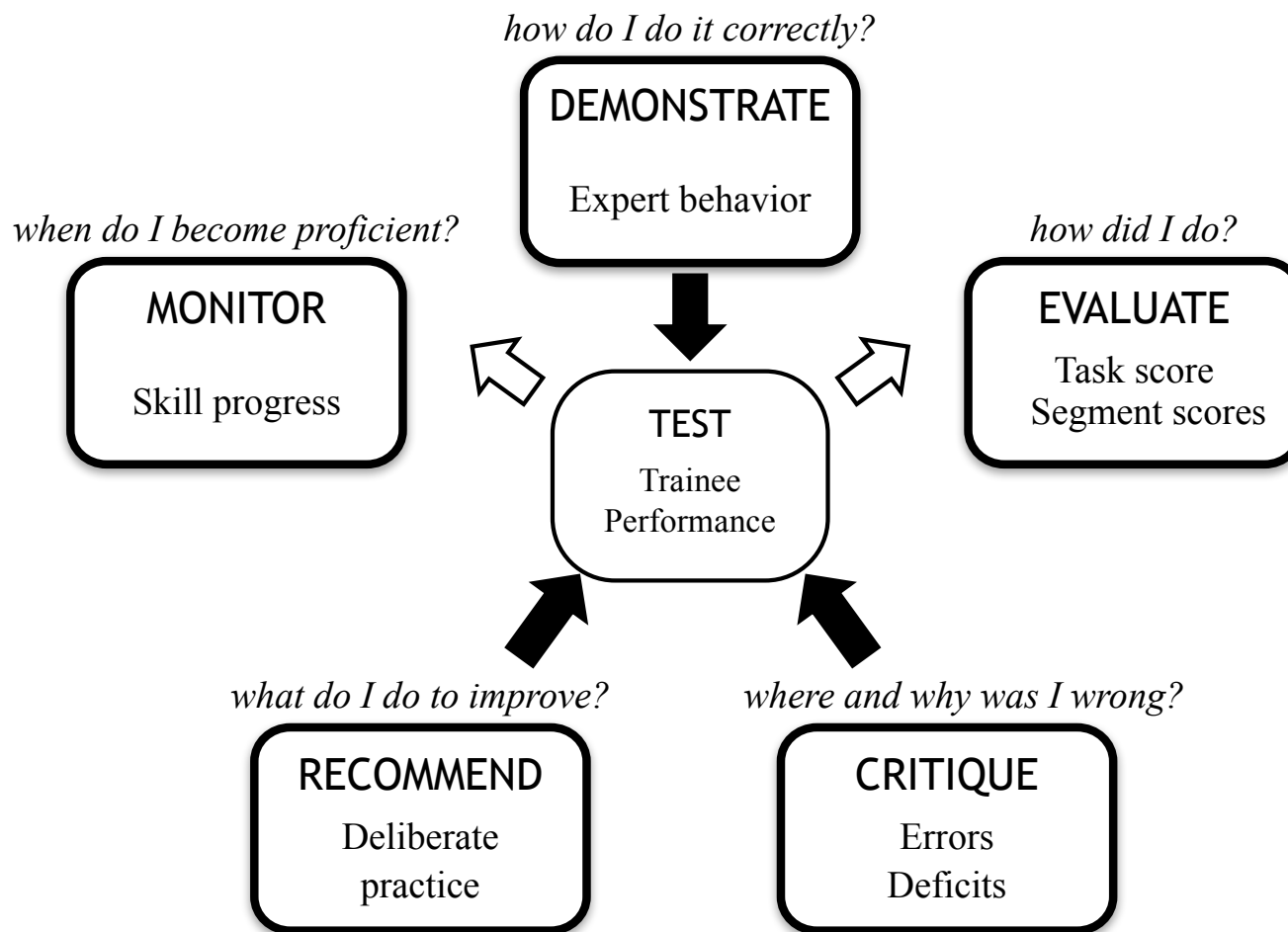
What Could You Do With a Million Surgeries?



Annual Worldwide Procedures (Figure 1)



Can Machines Help Train Surgeons?



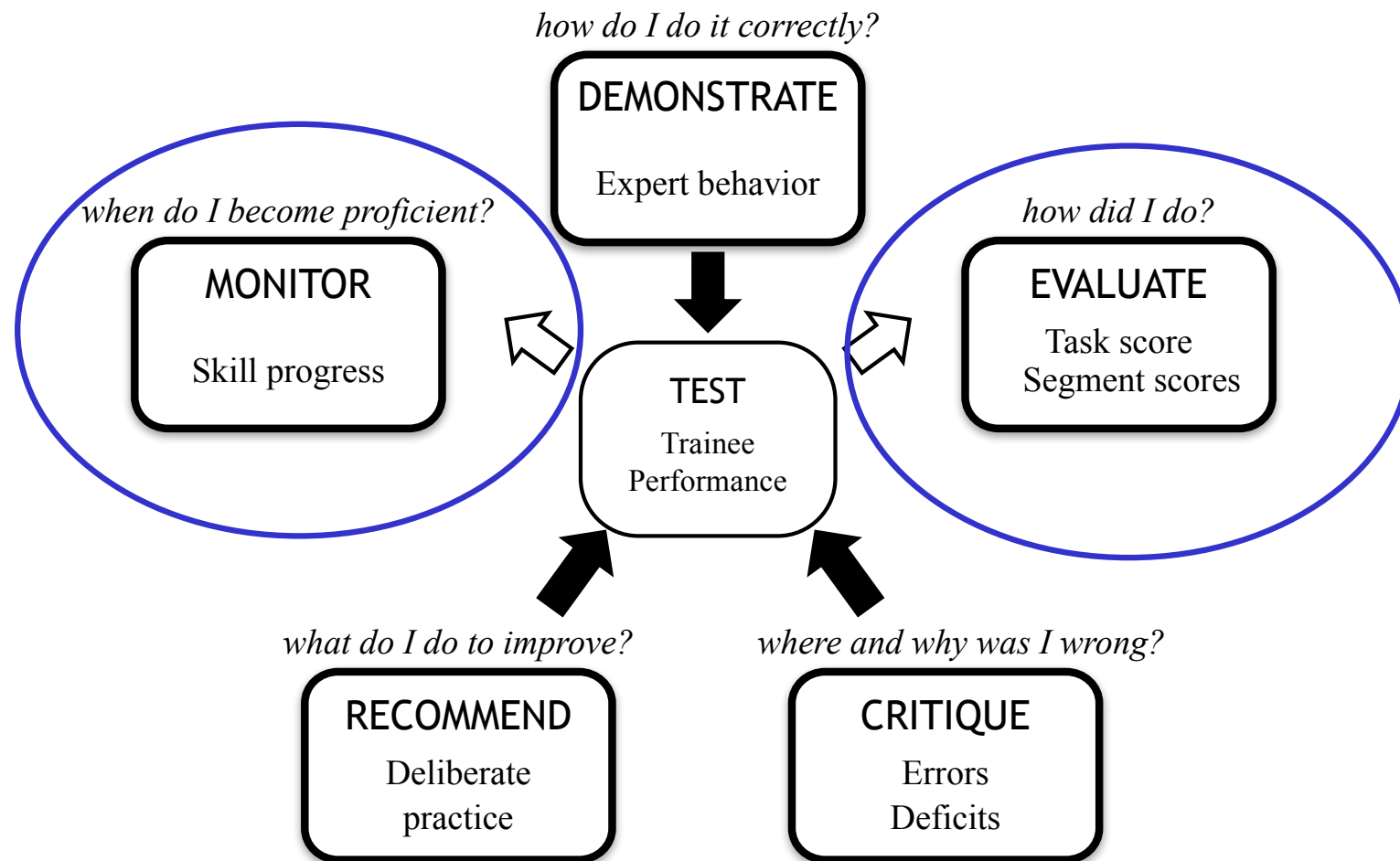
© Anand Malpani, 2016. Automated Virtual Coach for Surgical Training



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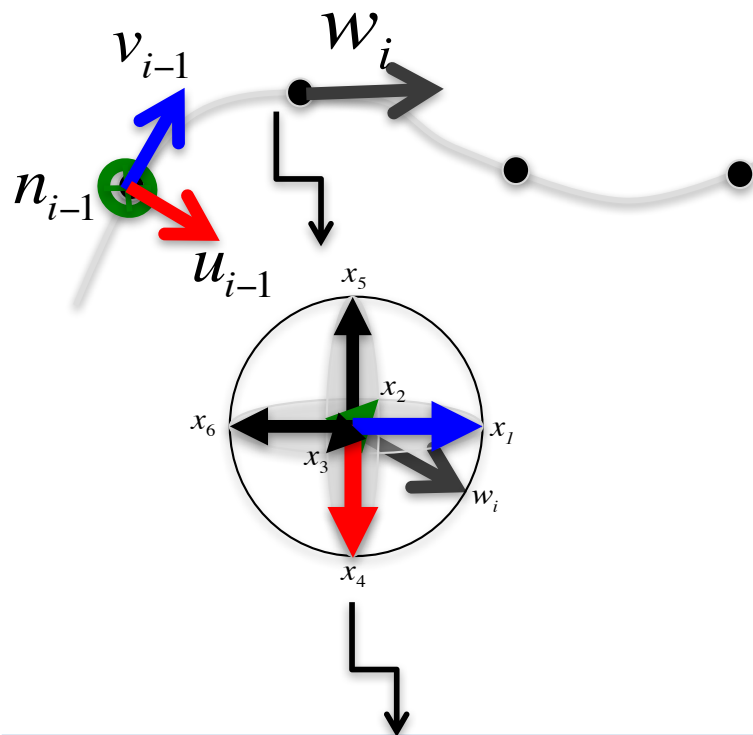
Can Machines Help Train Surgeons?



© Anand Malpani, 2016. Automated Virtual Coach for Surgical Training



Evaluate: How Am I Doing?



ID=1
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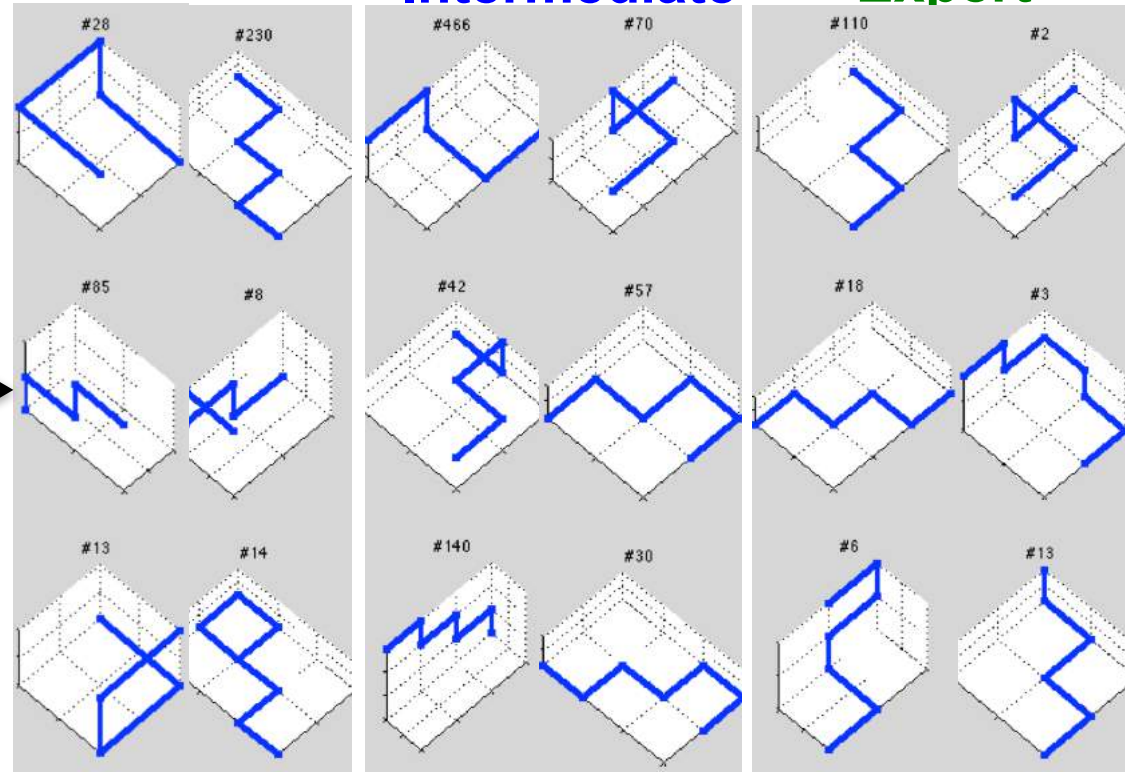
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ID=3
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Novice

Intermediate

Expert

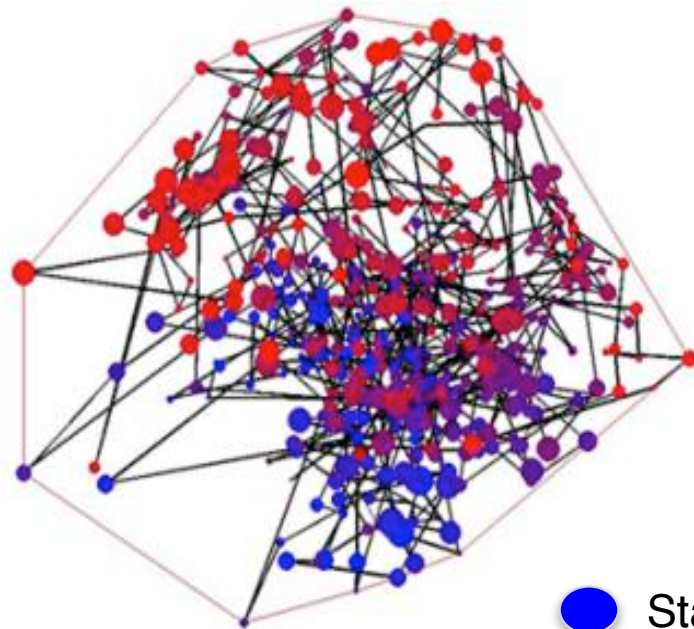
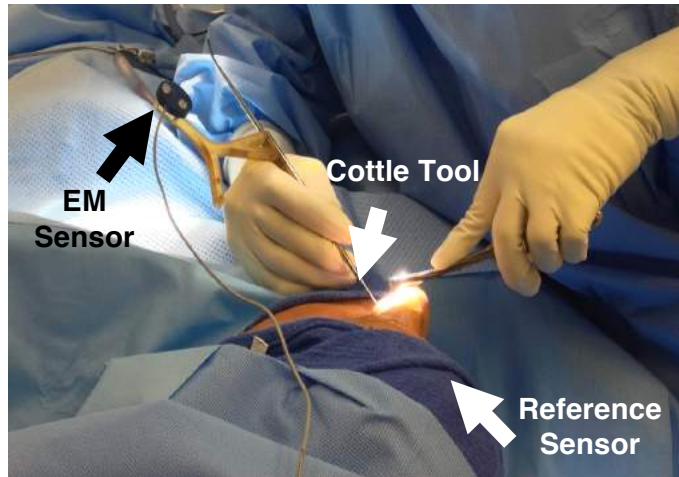


Translate movement to a string

Build category-specific dictionaries

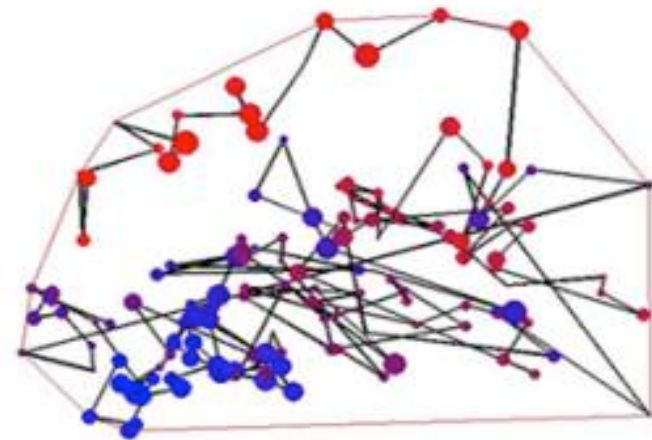
Ahmidi et al. Automated objective surgical skill assessment in the operating room from unstructured tool motion in septoplasty." IJCARS (2015)
Ahmidi, et al. "String Motif-Based Description of Tool Motion for Detecting Skill and Gestures in Robotic Surgery." MICCAI, 2013.

An Example: Sinus Surgery



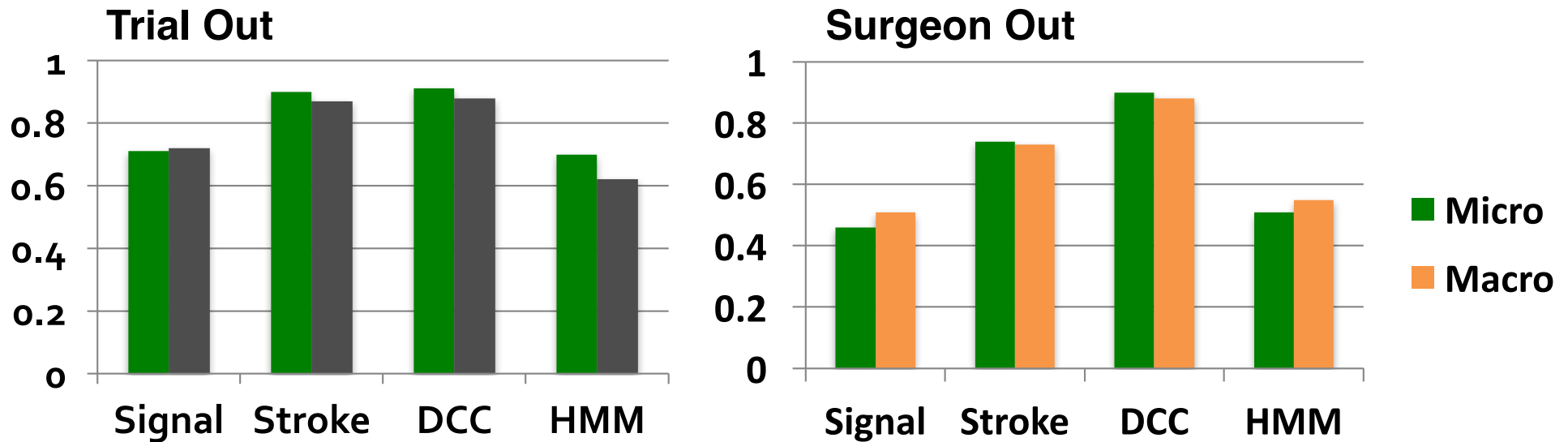
NOVICE

● Start of Surgery
● End of Surgery



EXPERT

Classifying Attending vs. Resident: Results



Ahmidi et al. Automated objective surgical skill assessment in the operating room from unstructured tool motion in septoplasty." IJCARS (2015)
[DCC] Ahmidi N, et al.: String Motif-Based Description of Tool Motion for Detecting Skill and Gestures in Robotic Surgery. MICCAI (2013)
[HMM] Varadarajan B, et al.: Data-derived models for segmentation with application to surgical assessment and training. MICCAI (2009)

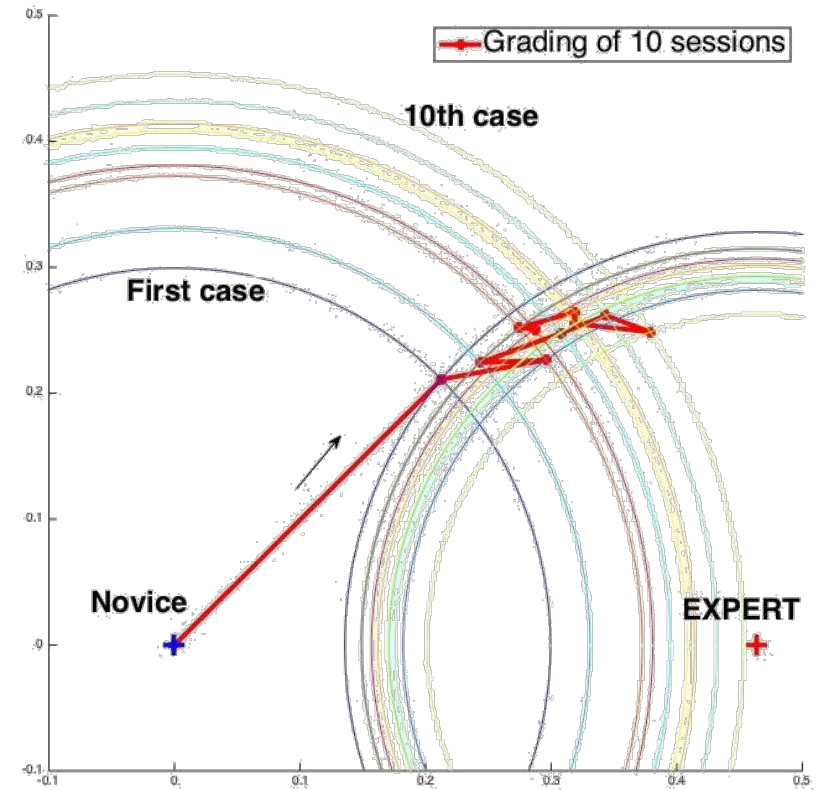
Monitoring Skill Progress



Providing On-Site Feedback

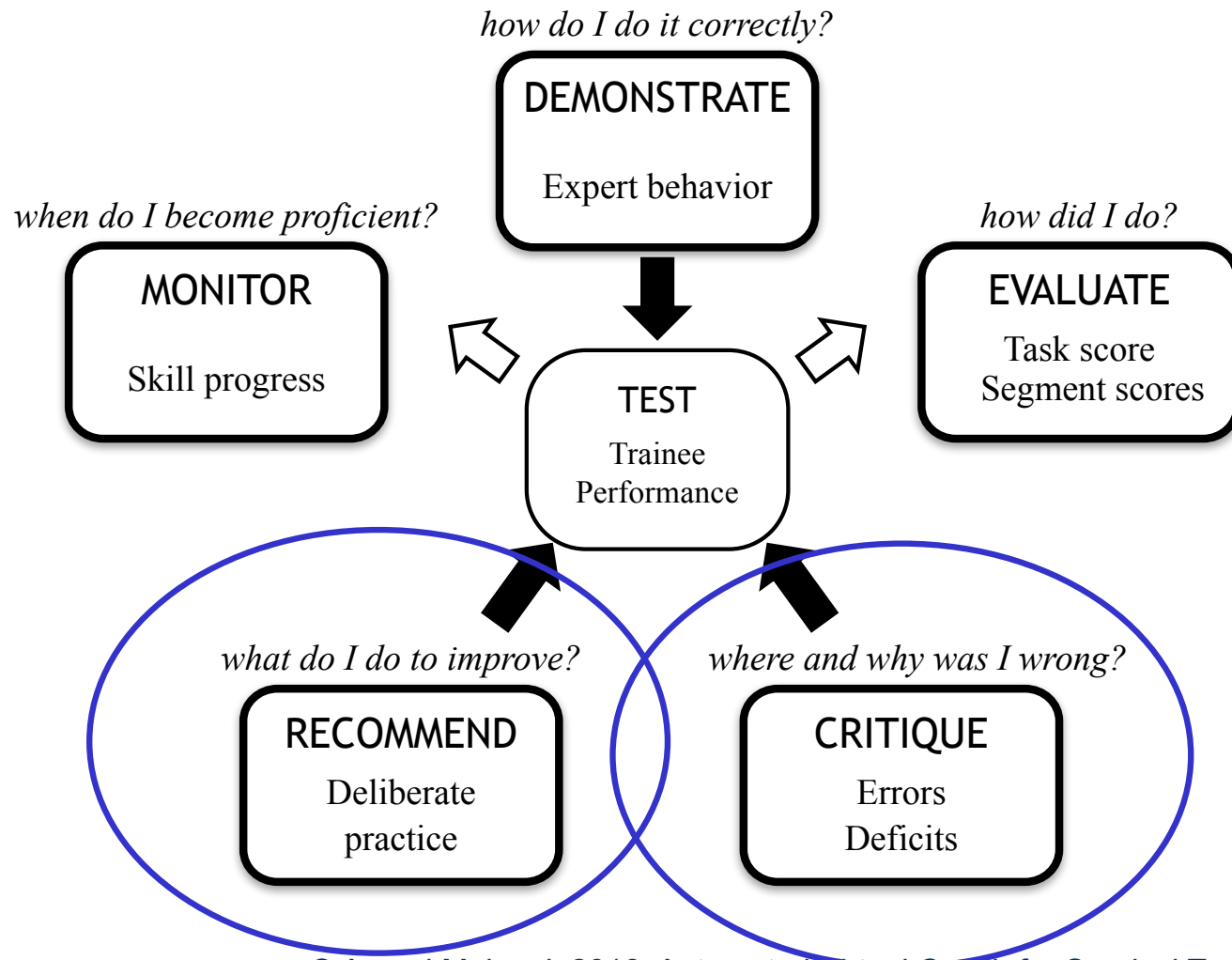
Collecting another 500+ trials (R01-DE025265 01)

- 2 academic medical centers
- 6 operating room suites
- 29 surgeons: 7 faculty; 22 trainees
- 181 procedures

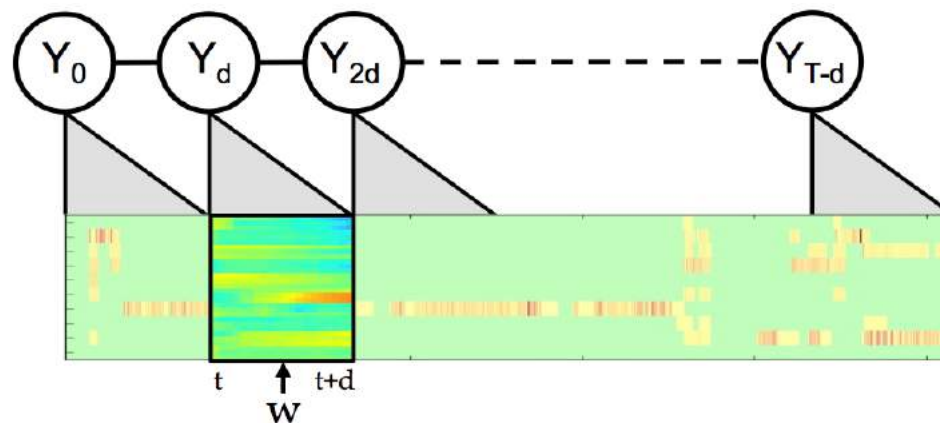
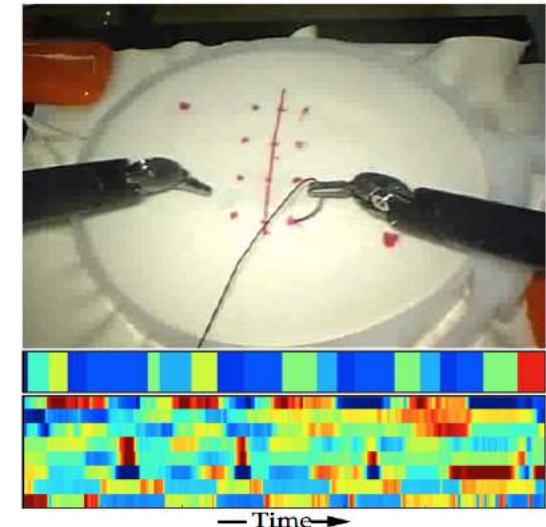
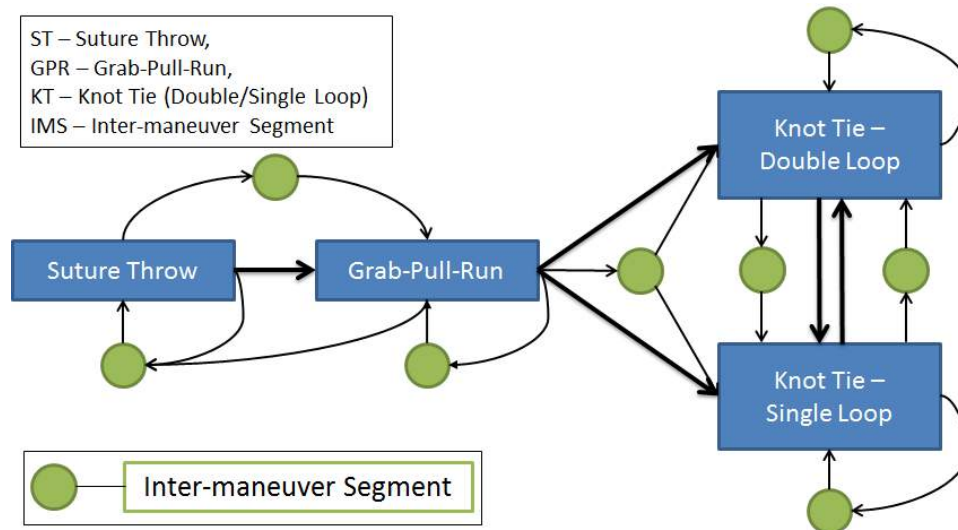


Following Trainees' Progress

Automated Feedback for Training



Where Did I Under-Perform?



Lea, Colin, et al. Temporal Convolutional Networks for Action Segmentation and Detection CVPR 2017.
 Lea et al. Learning Convolutional Action Primitives for Fine-grained Action Recognition ICRA 2016
 Lea et al. Segmental spatiotemporal CNNs for fine-grained action segmentation. ECCV 2016
 Lea et al. Surgical phase recognition: From instrumented ORs to hospitals around the world." M2CAI 2016.

Phases, Maneuvers, Gestures

	Spatial CNN			ST-CNN				
Data source(s)	LM	SMM	DTW	LM	SMM	DTW	[3]	[15]
Video	57.6	78.8	81.2	69.0	77.8	84.6	68.1	79.7*
Tools	58.5	76.5	85.7	56.4	78.3	91.2	78.9	73.0
Video + Tools	73.7	87.3	92.3	81.8	88.5	92.8	88.9	-

[3] Dergachyova IPCAI 2016 [15] Twinanda TMI 2016

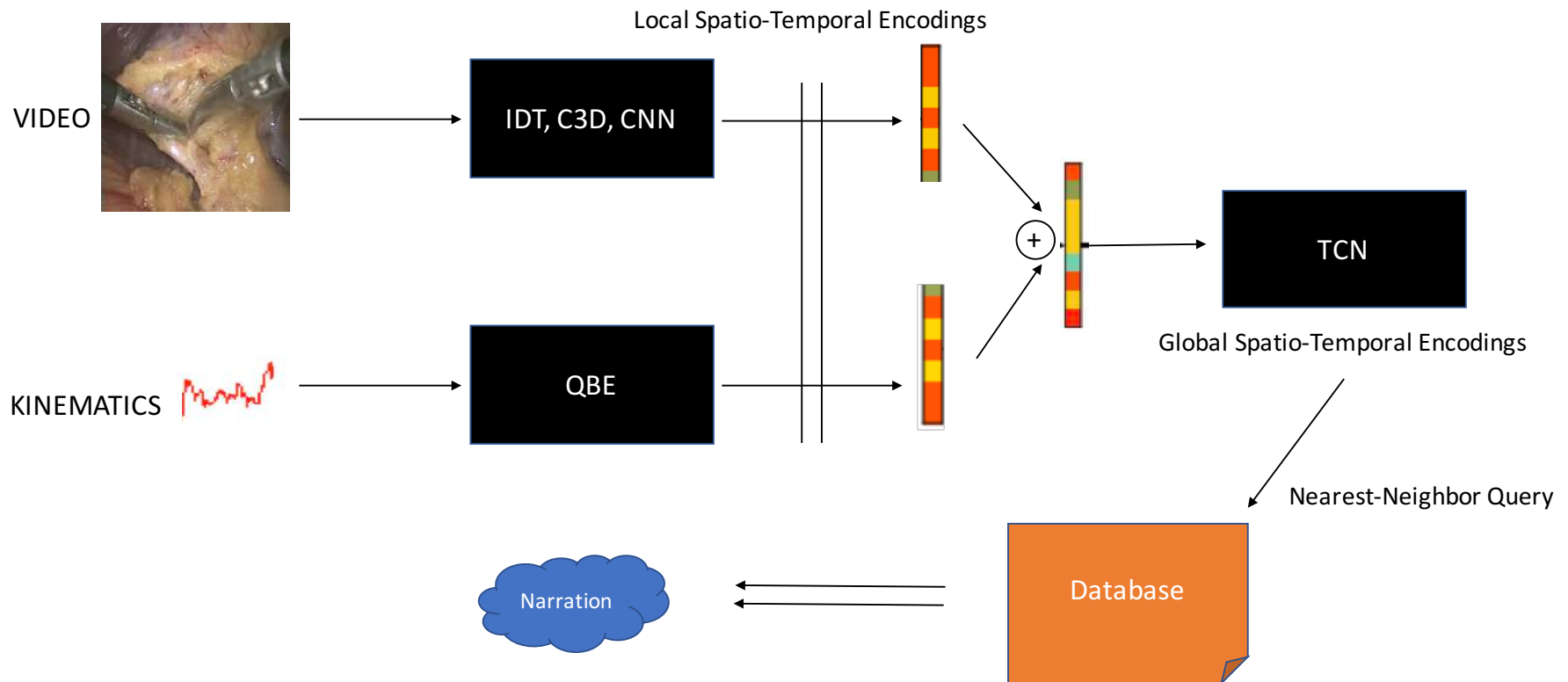
*=86.0% if trained using outside surgical data + tools

	JIGSAWS		MISTIC-SL	
	Accuracy (%)	Edit Dist. (%)	Accuracy (%)	Edit Dist. (%)
LC-SC-CRF [10]	82.5 ± 5.4	14.8 ± 9.4	81.7 ± 6.2	29.7 ± 6.8
Forward LSTM	80.5 ± 6.2	19.8 ± 8.7	87.8 ± 3.7	33.9 ± 13.3
Bidir. LSTM	83.3 ± 5.7	14.6 ± 9.6	89.5 ± 4.0	19.5 ± 5.2

Ahmidi et al. A dataset and benchmarks for segmentation and recognition of gestures in robotic surgery. *TBME*, 2017

DiPietro et al. "Recognizing surgical activities with recurrent neural networks." *MICCAI*, 2016

What Did I Do Wrong?

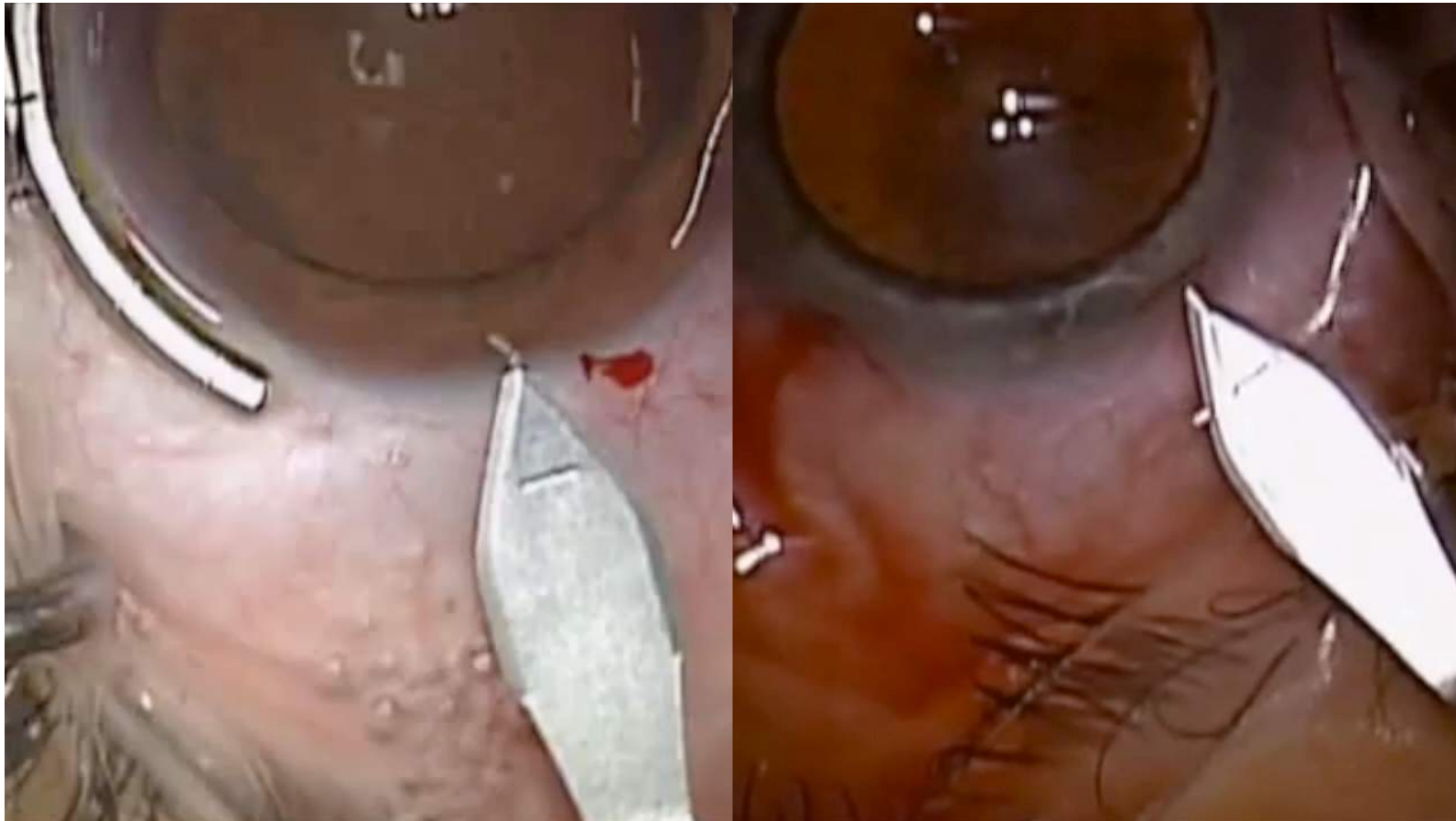


Use video as an index into feedback for a trainee based on most similar known performance

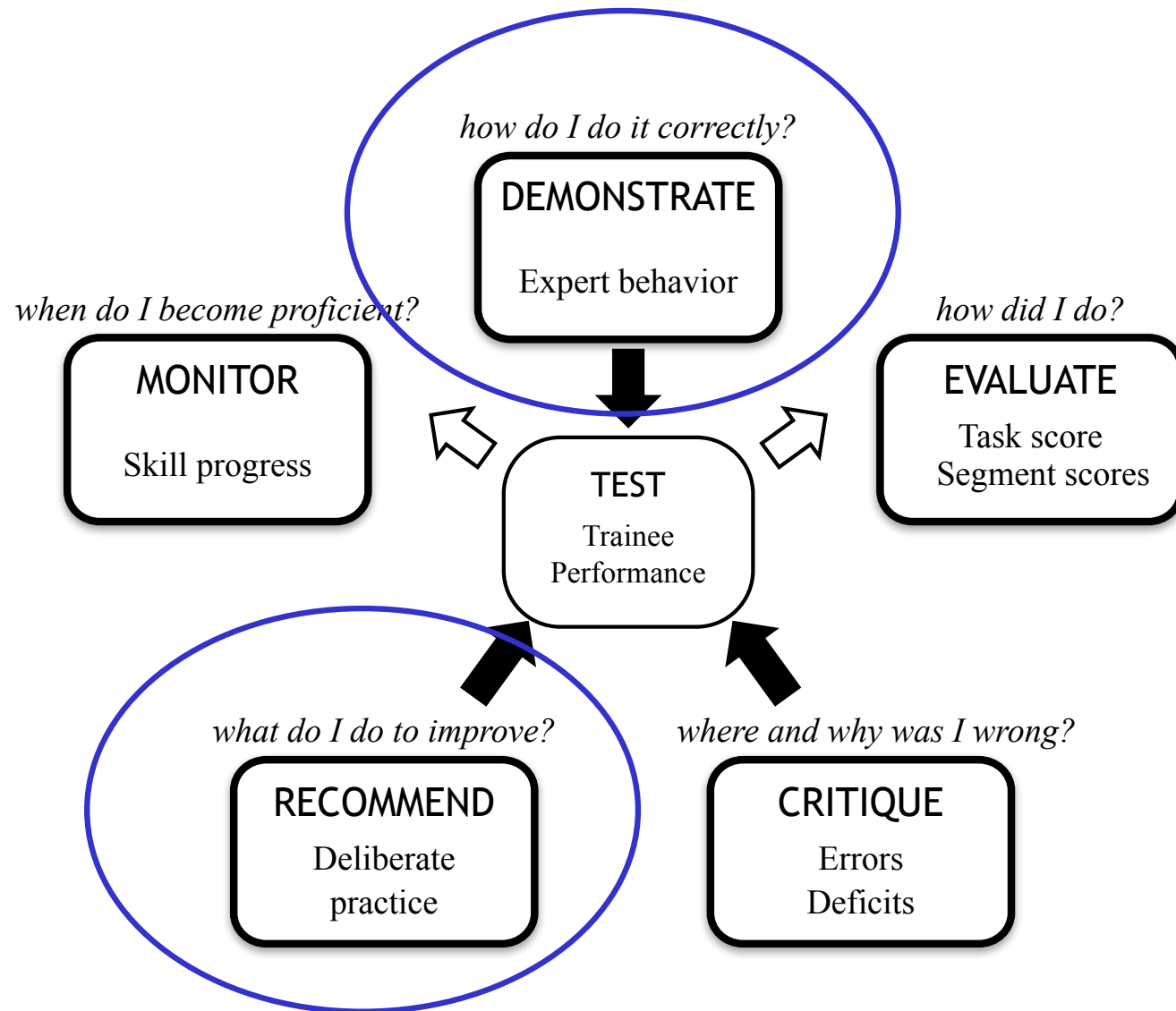
What Did I Do Wrong?

Query Video (Phase #2)

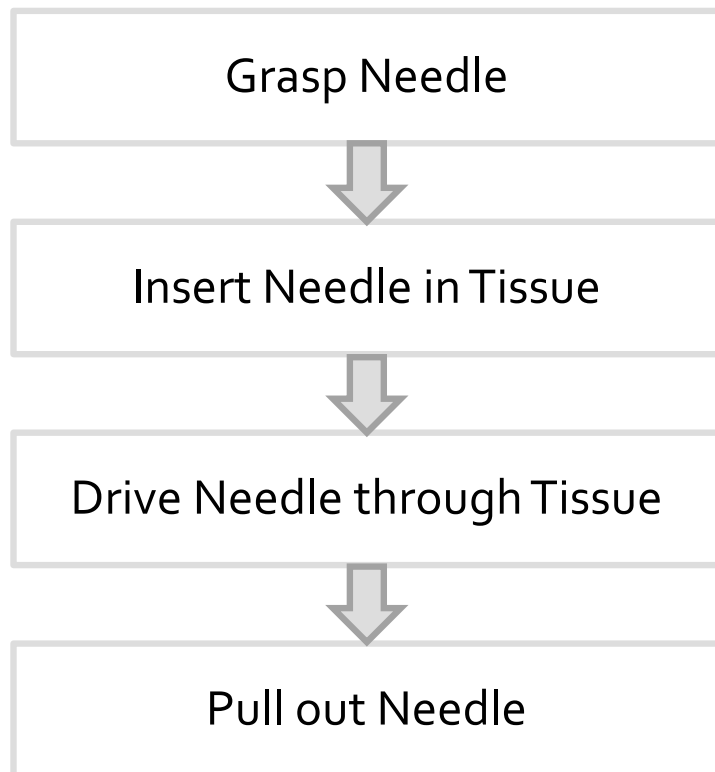
Closest Matching Clip



Automated Feedback for Training



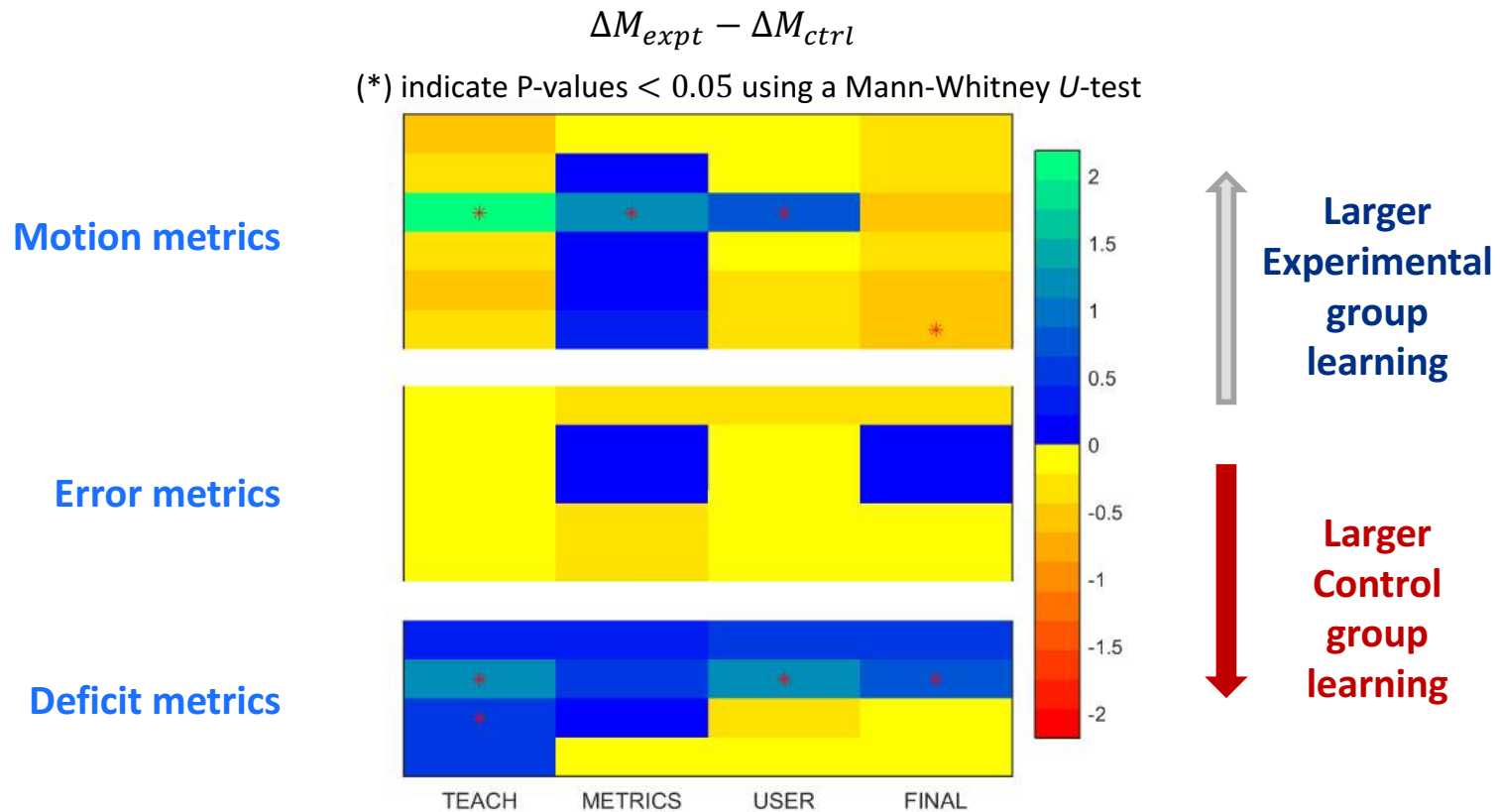
What Can I Do To Improve?



Malpani et al. Real-time Teaching Cues for Automated Surgical Coaching.

<https://arxiv.org/abs/1704.07436>

Improvement over baseline performance between groups



A Final Thought: Public Health Implications

- Number of Surgeries
 - US: 50M
 - Worldwide: 312M
- The Cost of Surgery
 - US: 180B

Rank	First-listed OR procedure*	Aggregate costs for hospital stays, \$ in millions	Percent of aggregate costs for stays with OR procedures, %	Mean cost per hospital stay, \$	Number of stays, in thousands
First-listed OR procedures		180,335	100.0	16,600	10,867
1	Spinal fusion	12,837	7.1	27,600	465
2	Arthroplasty of knee	11,317	6.3	15,900	711
3	Percutaneous coronary angioplasty (PTCA)	9,730	5.4	18,800	517
4	Hip replacement, total and partial	7,962	4.4	17,200	464
5	Cesarean section	7,481	4.1	5,900	1,269
6	Colorectal resection	6,747	3.7	23,400	289
7	Coronary artery bypass graft (CABG)	6,411	3.6	38,700	166
8	Heart valve procedures	6,070	3.4	53,400	114
9	Cholecystectomy and common duct exploration	5,048	2.8	12,600	400
10	Treatment, fracture or dislocation of hip and femur	4,275	2.4	16,800	255

Agency for Healthcare Research and Quality (AHRQ), Center for Delivery, Organization, and Markets, Healthcare Cost and Utilization Project (HCUP), Nationwide Inpatient Sample (NIS), 2011

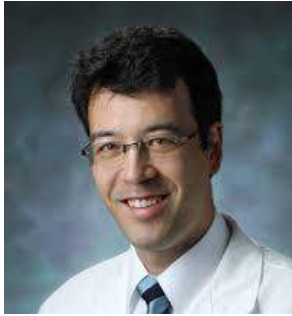
- Limitations
 - 5B people lack access to high quality surgical care
 - 5.6M deaths due to lack of access to quality surgical care
 - 9M procedures (3%) encounter major complications; 0.5% deaths

Surgical data science for next-generation interventions. Maier-Hein, Vedula,... Hager. *Nature Biomedical Engineering*, 2017
 Surgical data science: the new knowledge domain. Vedula, Hager. *Innovative Surgical Sciences*, 2017.



Machines can learn from people, not to replace them,
but to help make people better at what they do!

Clinical Collaborators



Masaru Ishii
Otolaryngology –
Head & Neck Surgery



Lisa Ishii
Otolaryngology –
Head & Neck Surgery



Mohamad Allaf
Urology



Shameema Sikder
Ophthalmology



Susan Gearhart
General Surgery



Gina Adrales
Minimally Invasive Surgery



Grace Chen
Obstetrics & Gynecology

Engineering Collaborators



Rene Vidal
BME



Sanjeev Khudanpur
ECE



Russ Taylor
CS



Austin Reiter
CS



Narges Ahmidi
Malone Center



Swaroop Vedula
Malone Center



Anand Malpani
Malone Center

The Motion Modeling and HMCS Mafia

Faculty:	Students and Staff:		Alumni:	
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Sanjeev Khudanpur, PhD	Colin Lea	Amanda Edwards	Amod Jog	Henry Lin, PhD
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Gyusung Lee, PhD	Lingling Tao	Chi Li	Balakrishnan Varadarajan, PhD	Nicolas Padoy, PhD
Lisa Ishii, MD	Mija Lee, PhD	Amir Ghalamazan	Balazs Vagvolgyi	Rajesh Kumar, PhD
Masaru Ishii, MD, PhD	Narges Ahmidi	Xiang Xiang	Benjamín Béjar, PhD	Sebastian Bodenstedt, MS
Michael Marohn, DO	Swaroop Vedula, MD PhD	Yixin Gao	Piyush Poddar, MS	Steven Hsiao, PhD
				Thomas Tantillo, MS

Publications: <https://publications.lcsr.jhu.edu/groups/HMM>

NSF NRI 1227277, NSF CPS 0931805, NSF CDI-II 0941362, NIH 1R21EB009143, NIH 5R21DE022656-02, NIH R01DE025265

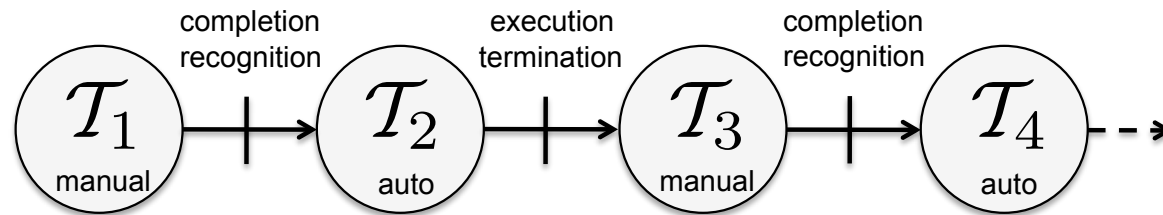
Intuitive Surgical, Inc., Science of Learning Institute (JHU)



QUESTIONS?



Learned Cooperative Execution



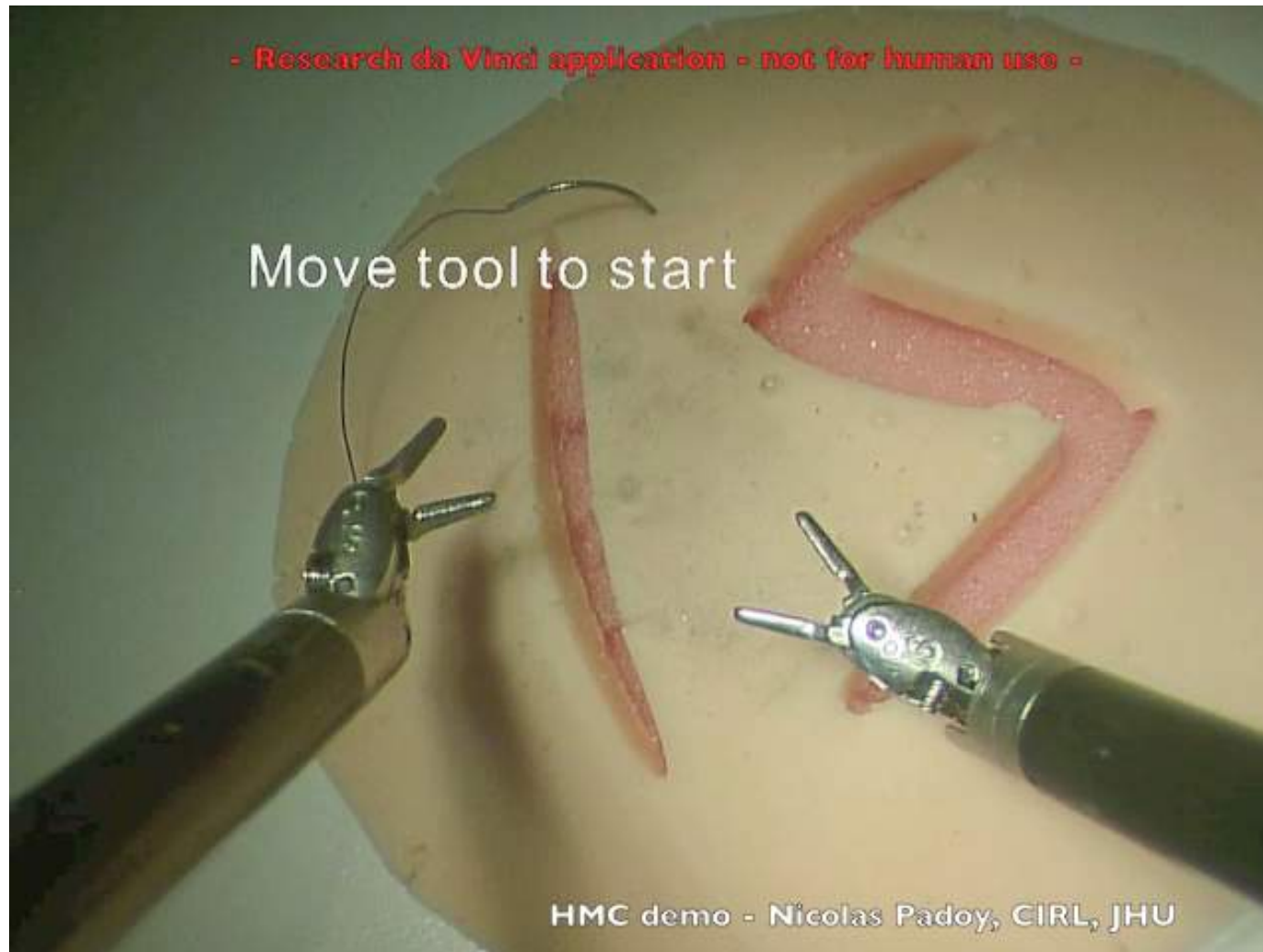
Suturing Task

#	Name
1	Grasp needle (RT) from pod, move to 1st suture point (RT), Insert needle (RT), grasp it (LT)
2*	Pull thread out (LT), move back to 2nd suture point (LT)
3	Grasp needle (RT) from (LT), Insert needle (RT), grasp it (LT)
4*	Pull thread out (LT), move back to 3rd suture point (LT)
5	Grasp needle (RT) from (LT) Insert needle (RT), grasp it (LT)
6*	Pull thread out (LT), move back to pod end point (LT)

Padoy and Hager, Human-Machine Collaborative surgery using learned models, ICRA 2011



From Data to Collaboration



Padoy and Hager. "Human-machine collaborative surgery using learned models." *ICRA 2011*

