

## 商用5Gネットワークを介した 国産手術支援ロボット"hinotori™" による遠隔手術の試み

神戸大学大学院医学研究科外科学講座 山口 雷藏

### 商用5G網を介した国産遠隔ロボット手術支援 プロジェクト概要

#### 遠隔医療の発展

社会実装

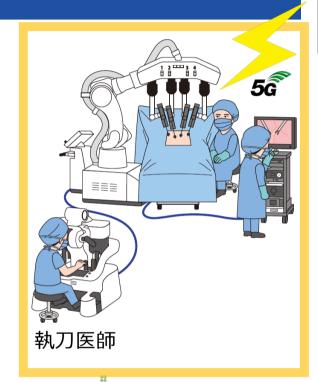
**遠隔ロボット手術支援への 臨床応用** 神戸大学

#### 発展のマネージメント

認定等の制度設計 ガイドラインの整備 業界団体の取り纏め 法整備の検討 拠点開発・企業誘致

#### 遠隔手術支援への応用

実証実験 臨床適用への条件設定 信頼性評価 市場リサーチ



光回線

((()))

A

#### **通信技術** NTT ドコモ

臨床使用に耐えうる 5Gネットワークの開発

### **医療ロボット技術** メディカロイド

通信ネットワークに接続できる 手術支援ロボットの開発



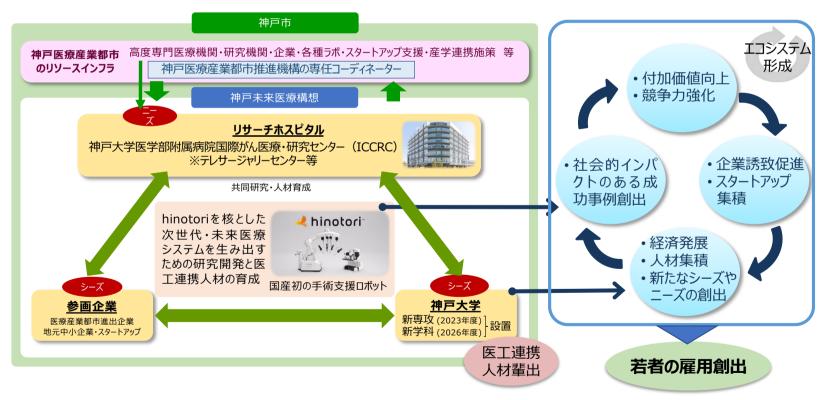




#### 神戸未来医療構想の事業概要

2020年度より内閣府、神戸市による地方大学・地域産業創生交付金事業として、

- ・研究環境整備事業:ICCRC、MeDIPを中心とした拠点整備
- ・研究開発事業:手術のデジタル化によるデーター医療産業の創出、遠隔ロボット手術の技術開発
- ・人材育成事業:医工連携人材の育成のための新専攻・新学科の設置



# 本日の内容

- 手術ロボット開発のキーワードはナビゲーション
- 手術ロボットとデジタル情報
- 遠隔ロボット手術は今
- ・最近の競合国の動き
- ・将来への発展性

### 今後のロボット支援手術の発展性

#### 3つのNavigation

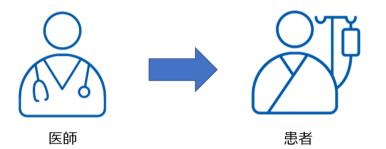
- ① がん細胞標識、分子イメージング、レーザー顕微鏡、8K内視鏡等によるハード 面からのナビゲーションアプローチ
- ② 集められた莫大な医療情報を**AI**等で解析し、手術技術の評価、指導といった ソフト面からのナビゲーションアプローチ
- ③ 5 Gによる機械間、病院間でのデータネットワークを確立することによって 様々な情報を開発センターへ収集するとともに、開発センターからの遠隔操作 等によるナビゲーションアプローチ



# 本日の内容

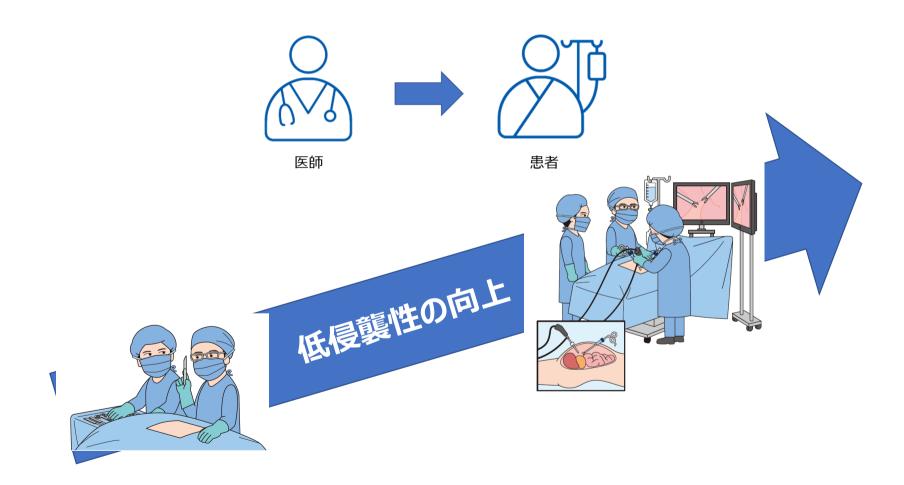
- 手術ロボット開発のキーワードはナビゲーション
- 手術ロボットとデジタル情報
- •遠隔ロボット手術は今
- ・最近の競合国の動き
- 将来への発展性

## 開腹手術

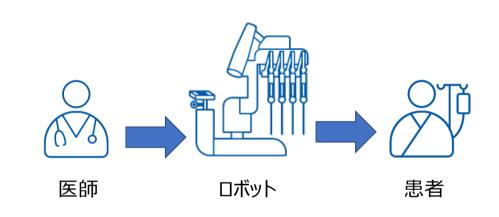




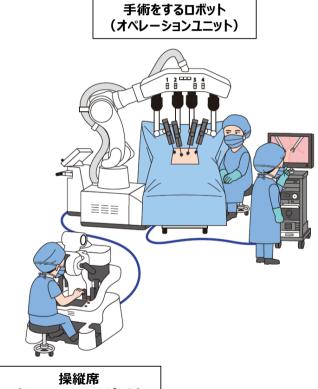
## 内視鏡下手術



## ロボット支援手術

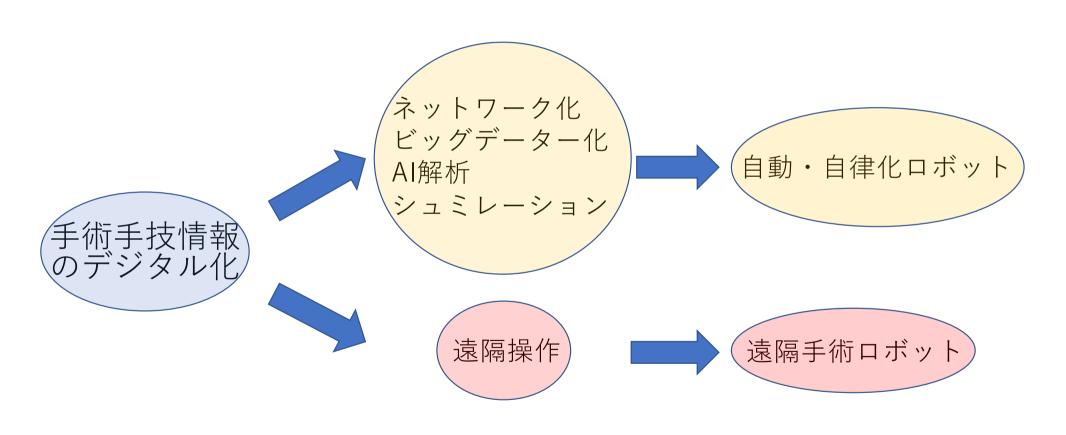


アナログからデジタルへ

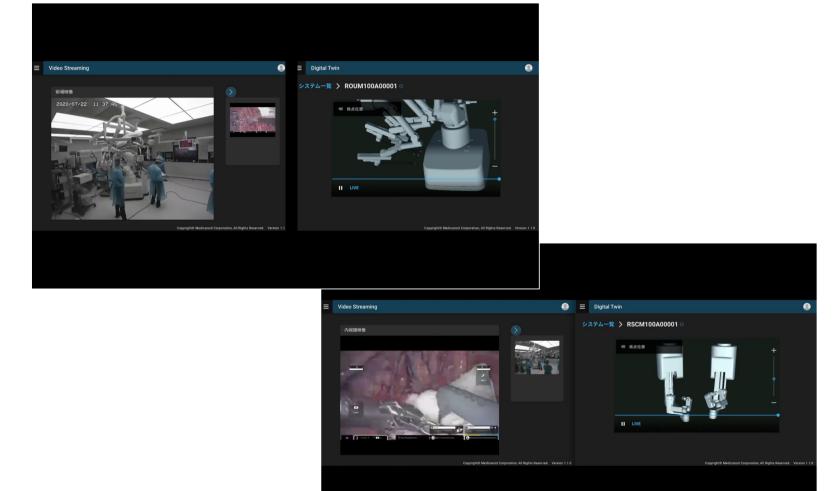


(サージョンコックピット)

## 手術手技のデジタル化がもたらしたもの



## ロボットの動作をリアルタイムに再現(Digital Twin)

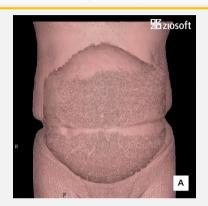


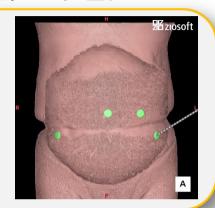


Copyright © 2020 Medicaroid Corporation All Rights R

## 術前手術計画~ポート配置シミュレーション~

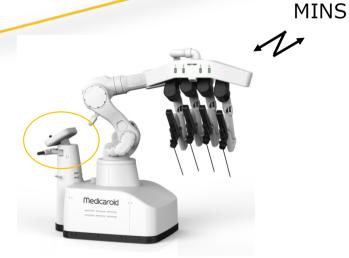






術前ポートシミュレーションを手術に反映

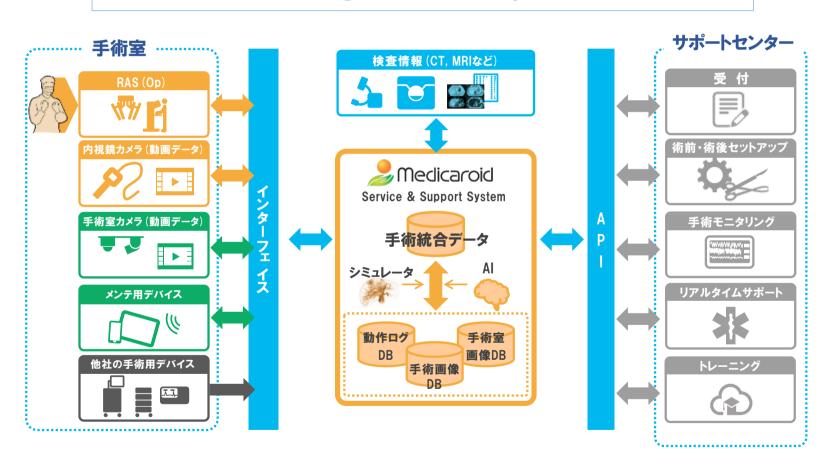




Copyright © 2020 Medicaroid Corporation All Rights Reserved.

## ネットワークサポートシステム

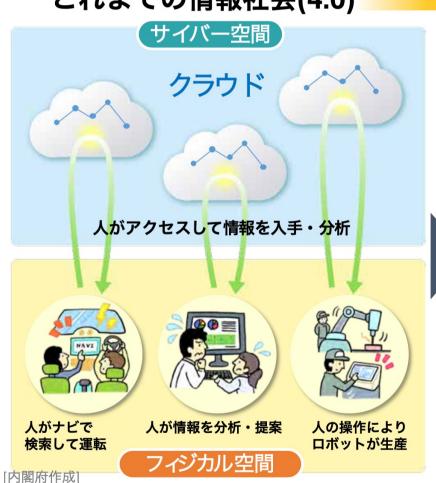
**Medicaroid Intelligent Network System (MINS)** 



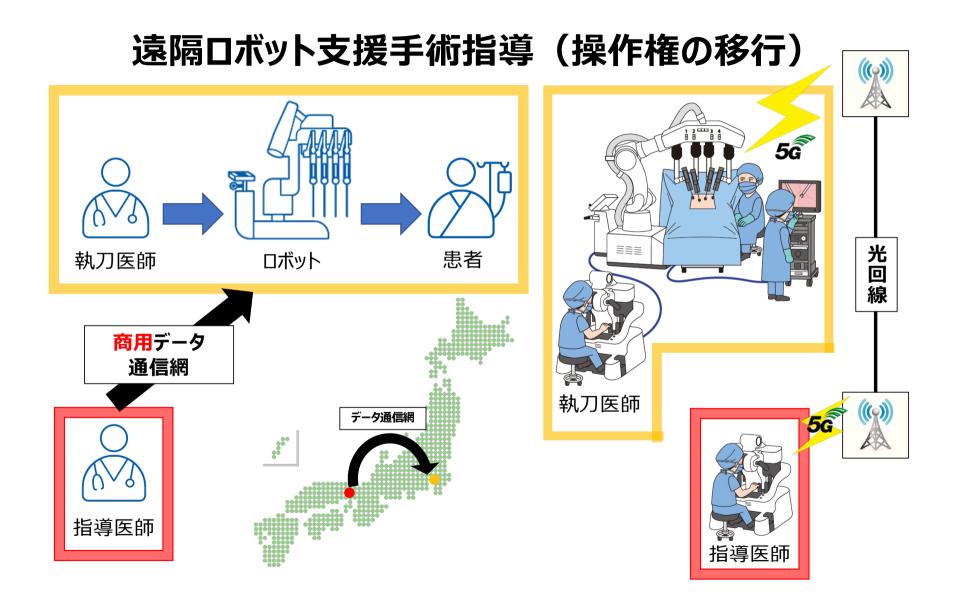
## サイバー空間とフィジカル空間の高度な融合

フィジカル(現実)空間からセンサーとLoTを通じてあらゆる情報が集積(ビッグデータ) 人工知能(AI)がビックデータを解析し、高付加価値を現実空間にフィードバック

### これまでの情報社会(4.0)



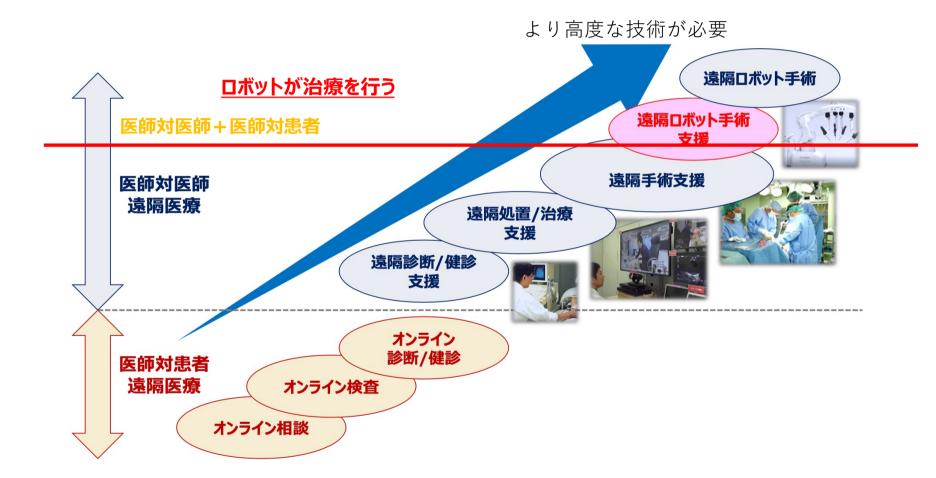
Society 5.0 サイバー空間 ビッグデータ 人工知能 5g センサー情報 高付加価値な情報、 環境情報、機器の作動情報、 提案、機器への指示など 人の情報などを収集 AIが人に提案 自動走行車で 工場で自動的に ロボットが生産 3 自動走行 ィジカル空間



# 本日の内容

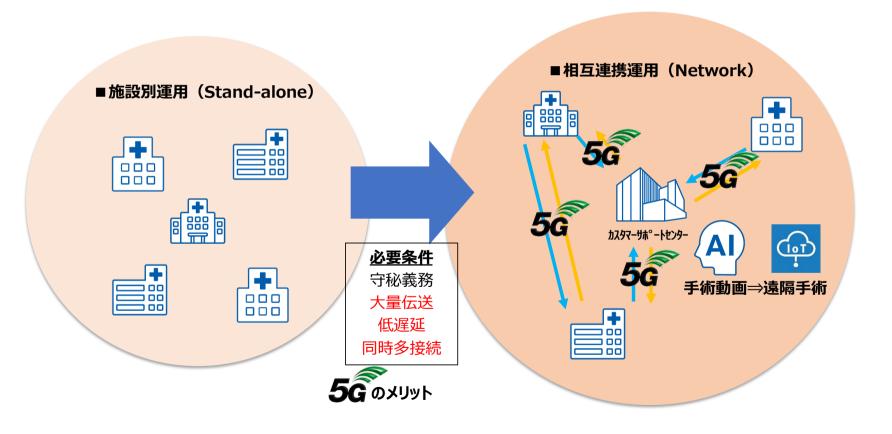
- 手術ロボット開発のキーワードはナビゲーション
- 手術ロボットとデジタル情報
- •遠隔ロボット手術は今
- ・最近の競合国の動き
- 将来への発展性

### モバイルネットワークを活用する遠隔医療



## デジタル通信の目指すところ

施設とロボット間、ロボット相互間をネットワーク化し、 医療情報の共有と遠隔でナビゲーションが行える仕組みの構築





#### **5G Telemedicine & Medical Training**

The purpose of this article is to inform about exciting activities and future visionary events taking place to enhance Department of Defense (DOD) medical support capabilities within the developing 5G core environment.

In June 2020, Office of the Undersecretary of Defense (Research & Engineering) named Joint Base San Antonio (JBSA) as an experimentation site for 5G augmented reality support for telemedicine and medical training. Although telemedicine is already happening today, it is often inhibited by a lack of adequate digital connectivity supporting the data speeds and volumes needed to provide real-time virtual healthcare. Since 5G is a critical strategic technology, the DOD must master 5G networks, which will eventually touch every mission and operation of DOD medicine.

The joint DOD telemedicine community has identified seven key goals that 5G solutions must address:

- Goal 1: Save lives and maximize preventive medicine.
- Goal 2: Provide resilient and fault-tolerant medicine and medical services anywhere in the world to support U.S. national interests and maintain military mission assurance.
- Goal 3: Collapse time and space to achieve real-time virtual and digital medical support.
- Goal 4: Extend medical expertise forward to the operational edge to enhance support for mobile forces in operational or austere environments.
- Goal 5: Ensure mission and data security in all medical applications.
- Goal 6: Maintain an environment which allows DOD medicine and medical services to practice virtual medicine to the maximum state possible.
- Goal 7: Provide an environment that facilitates the injection of new technologies in support of telemedicine applications.

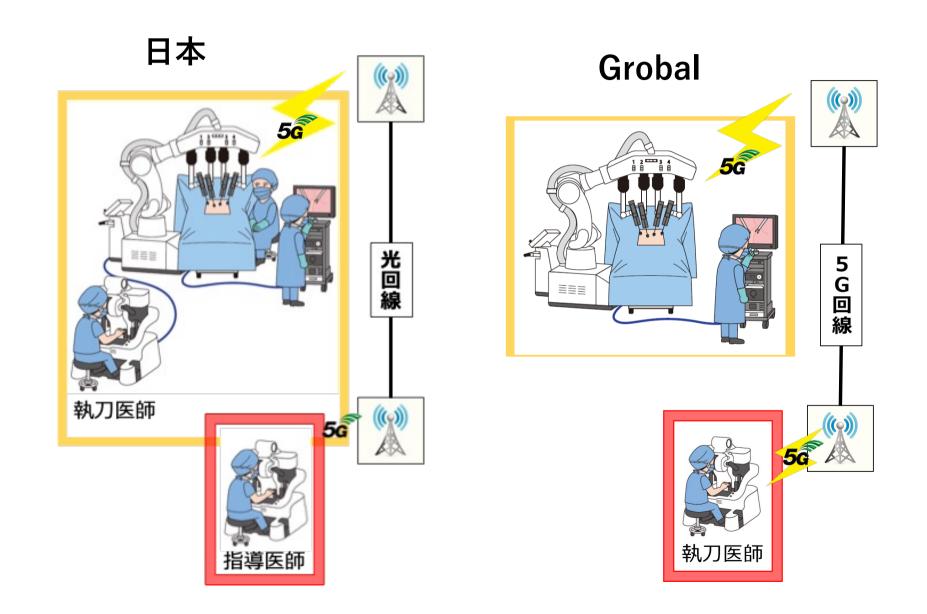
#### **5G Advanced Telerobotic Surgery**





# 本日の内容

- 手術ロボット開発のキーワードはナビゲーション
- 手術ロボットとデジタル情報
- •遠隔ロボット手術は今
- ・最近の競合国の動き
- ・将来への発展性





BJS, 2023, 110, 6-

https://doi.org/10.1093/bjs/znac27 Advance Access Publication Date: 23 August 207

Short Repo

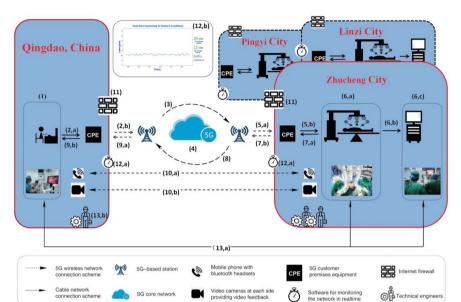
# Telerobot-assisted laparoscopic adrenalectomy: feasibility study

Jianmin Li<sup>1</sup>, Wei Jiao<sup>2</sup>, Hang Yuan<sup>2</sup>, Wei Feng<sup>3</sup>, Xuemei Ding<sup>4</sup>, Xulong Yin<sup>5</sup>, Liangjun Zhang<sup>6</sup>, Wei Lv<sup>7</sup>, Lufei Ma<sup>8</sup>, Liguo Sun<sup>8</sup>, Run Feng<sup>9</sup>, Jun Qin<sup>10</sup>, Xuefeng Zhang<sup>11</sup>, Chengyi Gou<sup>12</sup>, Shuxin Wang<sup>1</sup>, Zongyi Yu<sup>13</sup>, Bin Wei<sup>14</sup>, Lei Luo<sup>2</sup>, Fei Xie<sup>2</sup>, Yuan Chang<sup>15</sup>, Yonghua Wang<sup>2</sup>, Pier C. Giulianotti<sup>16</sup>, Qian Dong<sup>17</sup>, and Haitao Niu<sup>2</sup>, and Haitao Niu<sup>2</sup>, and Haitao Niu<sup>2</sup>, Pier C. Giulianotti<sup>16</sup>, Qian Dong<sup>17</sup>, and Haitao Niu<sup>2</sup>, and Hai

Table 1 Detailed intraoperative network communication and surgical outcomes of telerobot-assisted laparoscopic adrenalectomy

| Patient | Network communication |                         |                           |                                       | Surgical outcomes               |                    |                   |                         |                                |
|---------|-----------------------|-------------------------|---------------------------|---------------------------------------|---------------------------------|--------------------|-------------------|-------------------------|--------------------------------|
|         | Distance<br>(km)      | Network<br>latency (ms) | Total<br>latency<br>(ms)* | Network-<br>related adverse<br>events | Duration of teleoperation (min) | Blood<br>loss (ml) | NASA-TLX<br>score | Clavien-<br>Dindo grade | Pain score<br>at 24 h<br>(VAS) |
| 1       | 250                   | 54                      | 204                       | 2                                     | 60                              | 35.66              | 72                | I                       | 2                              |
| 2       | 82.5                  | 43                      | 193                       | 0                                     | 85                              | 68.47              | 72                | I                       | 1                              |
| 3       | 199                   | 30                      | 180                       | 1                                     | 58                              | 24.32              | 72                | I                       | 5                              |
| 4       | 199                   | 32                      | 182                       | 1                                     | 45                              | 4.85               | 97                | I                       | 3                              |
| 5       | 199                   | 29                      | 179                       | 0                                     | 73                              | 52.40              | 76                | I                       | 3                              |
| 6       | 82.5                  | 27                      | 177                       | 0                                     | 42                              | 17.69              | 43                | I                       | 1                              |
| 7       | 250                   | 35                      | 185                       | 0                                     | 38                              | 27.02              | 52                | I                       | 1                              |
| 8       | 250                   | 31                      | 181                       | 1                                     | 34                              | 22.32              | 90                | I                       | 1                              |
| 9       | 250                   | 30                      | 180                       | 0                                     | 49                              | 35.63              | 66                | I                       | 1                              |
| 10      | 250                   | 34                      | 184                       | 0                                     | 42                              | 26.20              | 67                | I                       | 1                              |
| 11      | 82.5                  | 22                      | 172                       | 0                                     | 44                              | 3.07               | 71                | I                       | 1                              |
| 12      | 82.5                  | 23                      | 173                       | 0                                     | 47                              | 20.30              | 72                | I                       | 2                              |
| 13      | 250                   | 32                      | 182                       | 1                                     | 70                              | 13.42              | 46                | I                       | 1                              |
| 14      | 250                   | 32                      | 182                       | 0                                     | 21                              | 10.35              | 44                | I                       | 1                              |
| 15      | 250                   | 34                      | 184                       | 0                                     | 21                              | 22.77              | 35                | I                       | 2                              |

\*The total latency (tl) includes network latency (nl) and the time for the robot system to process the signal (tr): tl = tr + nl. The tr was approximately 150 ms, including the servo period of the surgical robot (less than 1 ms), the mechanical response delay of the robot (40 ms), endoscopic imaging, image processing delay (50 ms), and the video codec delay (60 ms). NASA-TLX, National Aeronautics and Space Administration Task Load Index; VAS, visual analogue scale.



1 Equipment and network for telerobot-assisted laparoscopic adrenalectomy





In space, robots could someday make housecalls for doctors far from astronauts. Virtual Incision Corp. last week announced that it will test its MIRA surgical robot's skills in a 2024 technology demonstration mission aboard the International Space Station, or ISS.

The Robotics Applications Conference from Robotics 24/7.

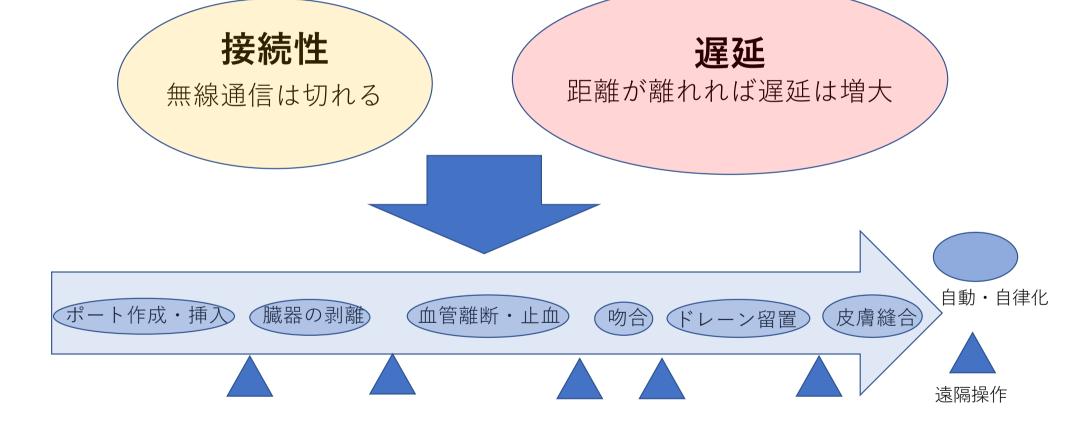
August10,2022

RAC Homepageより転載

# 本日の内容

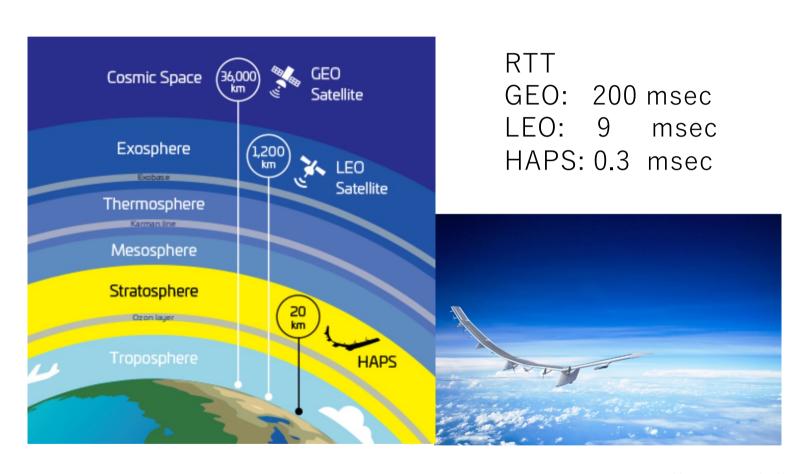
- 手術ロボット開発のキーワードはナビゲーション
- 手術ロボットとデジタル情報
- •遠隔ロボット手術は今
- ・最近の競合国の動き
- 将来への発展性

# 無線を介した遠隔手術の問題点

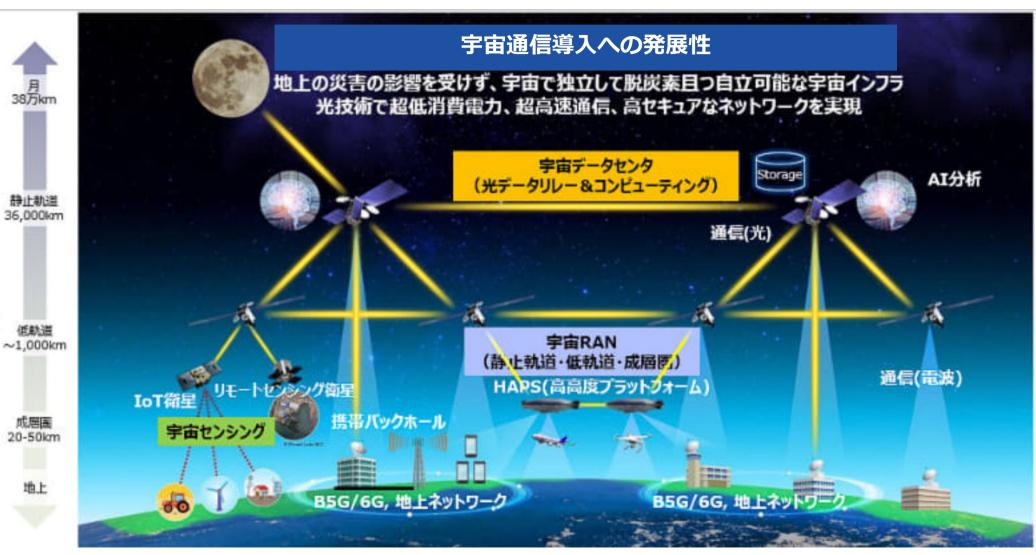


# 通信分野での発展性

## 無線通信の将来性



HAPS Mobile社のHPより転載



活動エリア

低軌道

地上

NTT Homepage 2022/4/26 ニュースリリースより







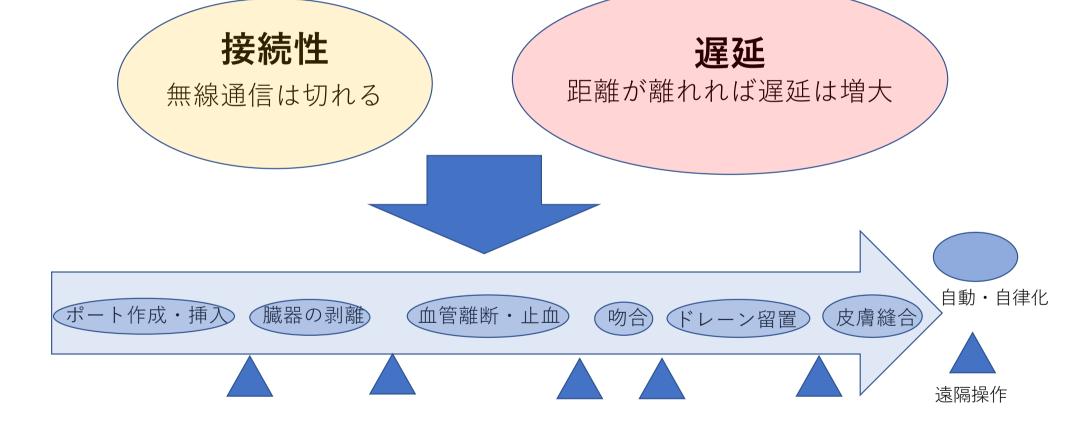


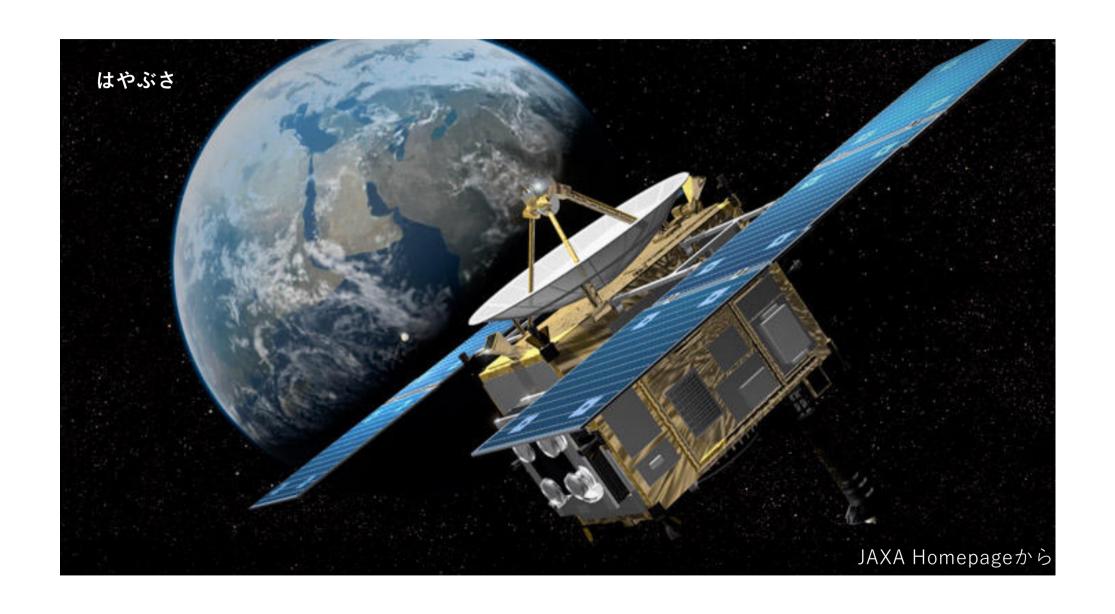




# ロボット分野での発展性

# 無線を介した遠隔手術の問題点

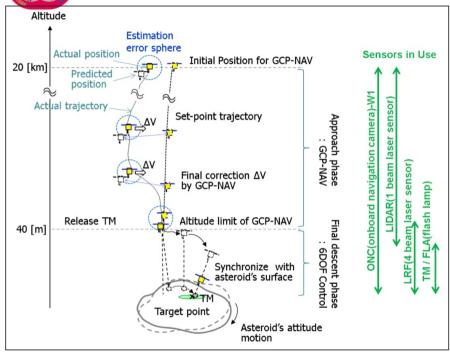




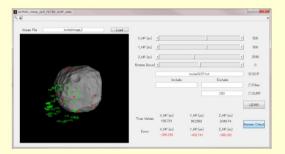


### JAXAのはやぶさ遠隔操作の技術を遠隔ロボット手術へ





- Ground Control Point Navigation (GCP-NAV)
- ✓ 20km~数100mまでの接近時に使用する遠隔操作 航法。
- ✓ 小惑星画像を地上に伝送。小惑星の特徴点、輪郭線をCGテンプレート画像とのマッチングで合わせこむことにより、探査機と小惑星の位置・姿勢情報を検出。
- ✓ これをもとに、エンジンの噴射量を地上で計算し探査機に指令を出す。
- ✓ 複雑な画像の認識、全体状況の瞬時判断は人間 が得意。通信時間遅れがあっても地上指示が有利。



GCP-NAV運用画面例

• Guidance Sequence Program (GSP)

- ✓ センサ情報に基づき、探査機が自律的に行う振る舞いのパタンを、地上から効率的に 書換え教示できるしくみ。
- ✓ 小惑星の表面状態さ光の反射度等、近傍観察し初めて得られる情報を把握した後、危 険判断の基準、危険時の対応等を地上運用者が分析の上決定し、自律動作を開始す る前に、地上指令として探査機内のテーブルを書き換える。
- ✓ 通信容量や探査機の計算機メモリの制約から効率的書換え・教示の仕組みが重要。

(© JAXA)

92

BJU Int 2019; **124:** 487–495 wileyonlinelibrary.com



# A deep-learning model using automated performance metrics and clinical features to predict urinary continence recovery after robot-assisted radical prostatectomy

Andrew J. Hung \*6, Jian Chen \*6, Saum Ghodoussipour\*, Paul J. Oh\*, Zequn Liu†, Jessica Nguyen\*, Sanjay Purushotham‡, Inderbir S. Gill\* and Yan Liu§

\*Center for Robotic Simulation and Education, USC Institute of Urology, Keck School of Medicine, University of Southern California, Los Angeles, CA, USA, \*School of Electronics Engineering and Computer Science, Peking University, Beijing, China, \*Department of Information Systems, University of Maryland, Baltimore, MD, and \*Computer Science Department, Viterbi School of Engineering, University of Southern California, Los Angeles, CA, USA

Table 1 Automated performance metrics and patient clinicopathological features.

#### APM Frequency of master clutch usage Time-related metrics Frequency of third arm swap Time to complete the task Frequency of energy application Moving time of the right instrument Number of times surgeon's head out of the console Moving time of the left instrument EndoWrist® articulation metrics Moving time of third instrument The total radians of the right instrument shaft rotation during the task Moving time of the camera The total radians of the right instrument wrist movement during the task Time of no instrument or camera movement The total radians of the right instrument jaw opening during the task Time of the right instrument not moving during the task The total radians of the left instrument shaft rotation during the task Time of the left instrument not moving during the task The total radians of the left instrument wrist movement during the task Time of the third instrument not moving during the task The total radians of the left instrument jaw opening during the task Time of the camera not moving during the task Right instrument articulation during the task Instrument kinematic metrics Left instrument articulation during the task Path length of the right instrument Angular velocity of the right instrument articulation Angular velocity of the left instrument articulation Moving velocity of the right instrument Clinicopathological features Path length of the left instrument Moving velocity of left instrument Path length of the third instrument BMI Path length of all three instruments Preoperative PSA Ratio of path length of right and left instruments Preoperative biopsy Gleason score Camera movement metrics Path length of the camera Surgery time Lymph node dissection template (standard vs extended) Moving velocity of the camera Number of camera adjustments during the task Urethropexy Frequency of camera adjustment Nerve-sparing Mean of time of each camera movement Prostatic median lobe Mean path length of each camera movement Final pathology Gleason score Mean of straight path length of each camera movement Pathological stage

Extracapsular extension

Prostate volume

Positive margins

Radiation received

100 cases included full follow-up data Step 1. Three prediction models were trained to Urinary continence recovery predict continence recovery after RARP prediction modeling and feature ranking Random survival Deep learning models Cox proportional based survival analysis hazards forests (DeepSurv) The mode with the highest accuracy (DeepSurv) was selected (Table 4) The top 5 ranked features were used to stratify the surgeons into two groups (Table 2) Step 1. Group 1/APMs Group 2/APMs Surgeon stratification and (More efficient APMs) (Less efficient APMs) historical cases comparison

161 RARPs from eight USC surgeons recorded

by a da Vinci system data recorder

Outcomes of 493 historical cases were

compared between the two groups (Table 6)

APM, automated performance metric; ASA, American Society of Anesthesiologists; BMI, body mass index.

System event metrics

Master clutch usage during task

Third arm swap during task

Energy usage during task

### **Objectives**

To predict urinary continence recovery after robot-assisted radical prostatectomy (RARP) using a deep learning (DL) model, which was then used to evaluate surgeon's historical patient outcomes.

### **Subjects and Methods**

Robotic surgical automated performance metrics (APMs) during RARP, and patient clinicopathological and continence data were captured prospectively from 100 contemporary RARPs. We used a DL model (DeepSurv) to predict postoperative urinary continence. Model features were ranked based on their importance in prediction. We stratified eight surgeons based on the five top-ranked features. The top four surgeons were categorized in 'Group 1/APMs', while the remaining four were categorized in 'Group 2/APMs'. A separate historical cohort of RARPs (January 2015 to August 2016) performed by these two surgeon groups was then used for comparison. Concordance index (C-index) and mean absolute error (MAE) were used to measure the model's prediction performance. Outcomes of historical cases were compared

using the Kruskal-Wallis, chi-squared and Fisher's exact tests.

#### **Results**

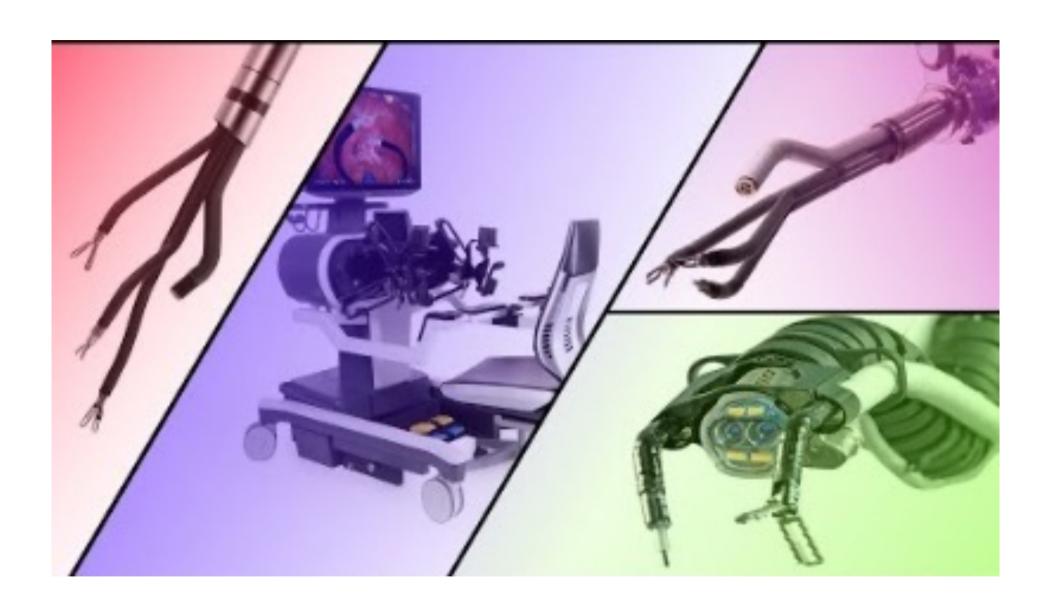
Continence was attained in 79 patients (79%) after a median of 126 days. The DL model achieved a C-index of 0.6 and an MAE of 85.9 in predicting continence. APMs were ranked higher by the model than clinicopathological features. In the historical cohort, patients in Group 1/APMs had superior rates of urinary continence at 3 and 6 months postoperatively (47.5 vs 36.7%, P = 0.034, and 68.3 vs 59.2%, P = 0.047, respectively).

#### Conclusion

Using APMs and clinicopathological data, the DeepSurv DL model was able to predict continence after RARP. In this feasibility study, surgeons with more efficient APMs achieved higher continence rates at 3 and 6 months after RARP.

### **Keywords**

robotic surgical procedures, prostatectomy, artificial intelligence, urinary incontinence, quality of life





ご清聴ありがとうございました