

# A novel development of deep neural network model for diagnosis of uterine sarcomas

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Uterine sarcomas are **rare**, occurring in approximately 5 of 10,000 women.

Bosch. et al. Best. Pract. Res. Clin. Obstet. Gynaecol, 2012.

The five-year overall survival rate **does not typically reach 50%**, especially among patients in the advanced stages.

Burghaus. et al. Arch. Gynecol. Obstet. 2016.

Uterine myomas with degeneration frequently mimic uterine sarcomas on MRI, and misdiagnosis of uterine sarcomas as benign myomas is not uncommon.

Sun. et al. *Diagn. Interv. Imaging.* 2019.

#### Introduction ~Image features of uterine sarcomas and myomas~



	Uterine myoma	Uterine sarcoma
Size	Variable	Variable (>10 cm, associated with poorer prognosis)
Margins	Well-defined	Irregular and ill-defined, often nodular and with invasion of adjacent structures
Signal on T1WI	Low to intermediate <b>High for fat content or hemorrhage</b>	Heterogenous and low High for hemorrhage from necrosis
Signal on T2WI	Generally, homogenous low signal <b>Intermediate/high in degeneration</b>	Intermediate to high signal
DWI and ADC signal	Low DWI, low ADC SI: ordinary Low DWI, high ADC SI: degenerated	Generally high DWI and low ADC SI
ADC values (10 <sup>-3</sup> mm <sup>2</sup> /s)	0.88-1.40	0.79-1.17
Contrast-enhanced MRI	Variable	Early heterogenous enhancement with central areas of contrast non-enhancement

SI: signal intensity, T1WI: T1-weighted images; T2WI: T2-weighted images.

Sun, *Diagnostic and Interventional Imaging*, 2019

Image features are **overlapped between uterine sarcomas and myomas**. In some cases, it is difficult to diagnose uterine tumors, especially **degenerated tumors**.

## Introduction ~Diagnosing problems of uterine tumors~





#### Accurate preoperative diagnosis is critical for patients.







#### **Artificial Intelligence (AI) : Reproduction of human's neural activity in a computer.**





#### **Training**



#### **Prediction**

Data: Unknown



Validation data set



#### Previous reports for imaging diagnosis of uterine sarcomas with AI



	Authors	Years	Article	Images	Sarcoma	Myoma	Accuracy	DNN / Machine learning
1	Malek et al.	2019	European journal of radiology	MRI	9	33	accuracy 91% AUC 0.972	Machine learning
2	Xie et al.	2019	European journal of radiology	MRI	29	49	accuracy 73.9% AUC 0.83	Machine learning
3	Nakagawa et al.	2019	Clinical Radiology	MRI PET	11	56	AUC 0.92	Machine learning
4	Lakhman et al.	2018	European radiology	MRI	19	22	accuracy 75%	Machine learning
5	Gerges L et al.	2018	American Journal of Roentgenology	MRI	17	51	sensitivity 88.2% specificity 78.4%	Machine learning
6	Wang et al.	2020	European Radiology	MRI	53	84	AUC 0.91	Machine learning
7	Malek et al.	2020	Scientific Reports	MRI	21	84	Accuracy 96.2% Sensitivity 100% Specificity 95%	Machine learning

No report which investigates diagnosis of uterine sarcoma using DNN.

\* Machine learning =conventional machine learning



**Objects of this study:** 

Investigation of **DNN model** for **imaging diagnosis of uterine sarcomas** 

**Our future goal:** 

Clinical application of our diagnosing DNN model in medical institutions



#### Methods ~study design~





## Methods: Patients and MR images





- Film-based MR imaging data
- Undergone pseudo-menopausal therapies.
- With other abdominal tumors (such as ovarian tumors and cysts).

15 types of MRI sequences	Abbreviation
Axial T1-weighted image	T1axi
Sagittal T1-weighted image	T1sag
Fat suppressed axial T1-weighted image	fsT1axi
Fat suppressed sagittal T1-weighted image	fsT1sag
Axial T2-weighted image	T2axi
Sagittal T2-weighted image	T2sag
Coronal T2-weighted image	T2cor
Fat suppressed axial T2-weighted image	fsT2axi
Diffusion-weighted image	DWI
Apparent diffusion coefficient map image	ADC
Axial dynamic contrast-enhanced image	dynamicaxi
Sagittal dynamic contrast-enhanced image	dynamicsag
Axial fat suppressed contrast-enhanced T1-weighted image	fsT1CEaxi
Sagittal fat suppressed contrast-enhanced T1-weighted image	fsT1CEsag
Coronal fat suppressed contrast-enhanced T1-weighted image	fsT1CEcor

### Methods ~ overall flow of this study ~





\*MobileNetV2

MobileNetV2 is one type of DNN network which consists of 88 layers and has around 3.5million learning parameters.

\*\*Ensemble prediction

Ensemble prediction is one method of machine learning and, in a simple term, a type of majority decision.

### **Results** ~ comparison of diagnosing accuracy ~





#### **Results of individual MRI sequences.**







#### **TOP10 of combination sets of MRI sequences.**

Combination set	<sup>1</sup> Accuracy	Sensitivity	Specificity	ADC	DWI	dynamicaxi	i dynamicsag	fsT1axi	fsT1CEaxi	fsT1CEcor	fsT1CEsag	fsT1sag	fsT2axi	T1axi	T1sag	T2axi	T2cor	T2sag
Set 1	91.3%	88.7%	94.0%															
Set 2	91.3%	89.8%	92.9%			•	٠	•	•	٠	•	•	•	•	•	•	●	•
Set 3	91.1%	91.9%	90.3%	•	•	•	•	•		•	•	•	•		•	•	•	•
Set 4	91.0%	87.8%	94.2%	•									•					•
Set 5	90.8%	90.7%	90.9%		•	٠	•	•	•	•		•	•		•	•	•	
Set 6	90.5%	90.5%	90.5%		•	•	٠	•	•	•		•	•		•	•	●	•
Set 7	90.5%	88.5%	92.5%	•	•	•	•	•	•	•	•		•		•	•	•	•
Set 8	90.5%	89.6%	91.3%	•		•	٠	•	•	•	•	•	•		•	•	•	
Set 9	90.4%	89.0%	91.8%		•	•	•		•	•	•	•	•	•	•	●	•	•
Set 10	90.3%	91.9%	88.8%		•	•	•	•	•	•	•	•	•	•		•	•	•
Average	90.8%	89.8%	91.7%															

The results **combining MRI sequences can be better** than that of individual sequences. **Axial T2WI, sagittal T2WI** and **DWI** seem to be important sequences for DNN models.

## **Results** ~ **AI-supported** examination~



		Accuracy	Sensitivity	Specificity
AI-supported	DNN model	90.8%	89.8%	91.7%
	Specialists	87.3%	83.1%	91.7%
	Practitioners	90.8%	87.8%	93.7%
Radiological Radiological specialists (n=3) practitioners (n=3)		The results of I	DNN model in the 1 <sup>st</sup>	examination were p

#### Comparison the average between 1st and 2nd examination.



DNN models can be helpful for diagnosing uterine sarcoma, especially **reducing occult tumors**, and **filling the gap** of readers skills.



#### **Our DNN model shows**

# **High diagnosing accuracy**

# Supporting ability to fill the gaps

# Supporting ability to reduce occult tumors



### Limitation of the first DNN model





### **Development of new DNN models**





## Overall flow of 3<sup>rd</sup> examination





### The evaluation for training data set





DNN models can recognize images containing tumors and keep high accuracy, which means new DNN model can automate preparation of MR images.

### The evaluation for validation data set





Our DNN models showed high diagnosing ability for the validation data set, which is important for clinical application.

### Conclusion



