

**Retour SNMMI 2021**

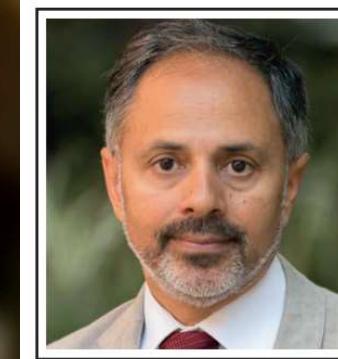
09/09/2021

A méditer ...

June 11-15

# SNMMI2021 *Virtual* ANNUAL MEETING

*"Innovation is not always a new device or technology, but simply finding a new/novel way to use an existing technology."*



## A Tribute to Sanjiv (Sam) Gambhir

The molecular imaging community has lost a brilliant scientist and visionary, but, more importantly, a genuinely empathetic and caring individual and a friend. We are profoundly grateful for his vast contributions to science, medicine and humanity. Our thoughts are with his wife, Aruna, and the extended Gambhir family.

# Comment utiliser au maximum les informations disponibles ?

## SUVmax described in FDG PET-CT reports can provide information of lesion location: an investigation of real-world data

Kenji Hirata, Yuko Uchiyama, Shiro Watanabe, Sho Furuya and Kohsuke Kudo

Journal of Nuclear Medicine May 2021, 62 (supplement 1) 1194;

Background	Methods	Results																																																			
<ul style="list-style-type: none"><li>FDG-PET/CT reports often describe the maximum standardized uptake value (SUVmax) to express the intensity of FDG uptake.</li><li>SUVmax is written with various decimal places (DP):<ul style="list-style-type: none"><li>1st DP 3.1 (indicating 3.05 ~ 3.15)</li><li>2nd DP 3.14 (indicating 3.135 ~ 3.145)</li><li>3rd DP 3.142 (indicating 3.1415 ~ 3.1425)</li></ul></li><li>If SUVmax is written in higher decimal places, it could have information for identifying the lesion location in the image, because there should be limited number of voxels that satisfy the SUVmax description.</li><li>We published preliminary data (Hirata et al. <i>Front. Med.</i>, <u>2021</u>), demonstrating that 97.8% of lesions could be uniquely identified under the condition that 1) SUVmax &gt; 5, 2) SUVmax is described with 3rd DP, and 3) local maximum restriction is applied.</li><li>However, since the <b>real-word</b> PET/CT reports were not used in the previous study, it has not been clear how well the method works for the actual reports.</li><li><b>Aim of the study:</b> To investigate the FDG-PET/CT reports in our institute retrospectively to clarify whether SUVmax in the reports can be used as an identifier for lesions.</li></ul>	<ul style="list-style-type: none"><li>Retrospective single-center study</li><li>Hokkaido University Hospital IRB approval (No.017-0454)</li><li>Informed consent waived</li><li>Study period: a total of 20 working days from the periods of January 11-20, 2017, May 11-20, 2014, and September 11-20, 2011.</li><li>A total of 230 FDG-PET/CT examinations were investigated.</li><li><b>Regular expression</b> was used to extract the strings that possibly represented SUVmax in the PET/CT reports, a regular expression formula <math>\\$\text{d}+(\.\text{d})\\$\text{d}+</math> was used to extract float numbers (e.g., 3.14).</li><li><b>Post-processing</b> after the regular expression process, the followings were excluded:<ul style="list-style-type: none"><li>containing 2 period characters (e.g., 2020.11.6) (date string)</li><li>"MBq" followed (e.g., 234.5MBq) (dosage string)</li><li>"Previous study" preceded within 4 characters.</li></ul></li><li><b>Radiologist's evaluation:</b> an experienced nuclear medicine physician evaluated all the SUVmax strings to determine true SUVmax or not.</li><li><b>The algorithm for lesion localization based on SUVmax string</b><ul style="list-style-type: none"><li>All voxels in the whole-body image were searched to extract the voxels having a nearby value.</li><li>In case of SUVmax = 3.14, voxels of <math>3.135 \times 3.145</math> were extracted.</li><li>Local maximum restriction: only the voxel that was local maximum within the <math>5 \times 5 \times 5</math> voxel space.</li><li>In case only 1 voxel satisfied the criteria, it was defined as <i>successful</i>.</li><li>In the example case, the voxel 8.1 is local maximum, but 7.1 is not.</li></ul></li><li>Software environments<ul style="list-style-type: none"><li>Visual Studio 2019 version 16.7.5</li><li>C# &amp; .NET Core 3.1</li><li><b>Metavol</b></li><li>Fellow Oak DICOM</li><li>Math.NET</li></ul></li></ul>	<p><b>Case 1:</b> Recurrence of rectal cancer. The PET/CT report indicated following lesions:</p> <ul style="list-style-type: none"><li>Right inguinal (SUVmax=6.482)</li><li>Pancreatic body (9.648)</li><li>Right spermatic cord (4.546)</li></ul> <p>SUVmax was written with 3rd DP. All the lesions were appropriately identified using the proposal method</p> <p><b>Case 2:</b> Lung cancer. The PET/CT report indicated following lesions:</p> <ul style="list-style-type: none"><li>Right upper lobe of the lung (SUVmax=18.2)</li><li>Left adrenal gland (6.3) <i>Detection Failed</i></li><li>Para-aortic node (6.6)</li></ul> <p>SUVmax was written with 1st DP. Two SUVmax strings were appropriately associated to the voxel, but the other one was not. "SUVmax=6.3" corresponded to 3 different voxels, among which one (arrow) was the true lesion.</p> <p>435 SUVmax strings were automatically detected in 230 PET/CT exams.</p> <p>After physician's inspection, 333 strings were judged as true SUVmax strings (i.e., 102 strings were excluded).</p> <p>The success rate varied depending on (1) the SUVmax range of the lesion and (2) the precision of the SUVmax string.</p> <p>In case <math>20 \leq \text{SUVmax}</math>, 91% cases were successful, whereas only 13% cases were successful in case <math>\text{SUVmax} &lt; 2</math>.</p> <p>When SUVmax was written with 1st decimal places (DP), only 29% cases were successful. In contrast, 91% cases were successful when written with 3rd DP.</p>																																																			
<p>Question: Is 1:1 correspondence possible between report and image using SUVmax?</p> <p><b>PET/CT report</b></p> <p>An uptake mass (SUVmax=5.093) is observed in the mediastinum, indicating paratracheal lymph node metastasis....</p> <p><b>Conclusion</b></p> <p>SUVmax strings in the FDG-PET/CT reports could be used for lesion localization. Further algorithms will improve the performance.</p>	<p><b>Example case (number = SUV)</b></p> <table border="1"><tr><td>7.9</td><td>6.1</td><td>8.0</td><td>7.7</td></tr><tr><td>6.5</td><td><b>8.1</b></td><td>5.5</td><td>6.5</td></tr><tr><td>7.4</td><td>5.8</td><td><b>7.1</b></td><td>6.2</td></tr><tr><td>7.5</td><td>5.9</td><td>6.9</td><td>5.8</td></tr></table> <p><b>Disclosures</b> The authors have no conflicts of interest with regard to this research.</p> <p><b>Funding</b> This work was supported by JST COI Grant Number JPMIC1301 (H31W09) and JSPS KAKENHI Grant Number JP20K08015.</p>	7.9	6.1	8.0	7.7	6.5	<b>8.1</b>	5.5	6.5	7.4	5.8	<b>7.1</b>	6.2	7.5	5.9	6.9	5.8	<p><b>Table: Success rate by SUVmax range and decimal places (DP)</b></p> <table border="1"><thead><tr><th>SUVmax</th><th>1<sup>st</sup> DP</th><th>2<sup>nd</sup> DP</th><th>3<sup>rd</sup> DP</th><th>Total</th></tr></thead><tbody><tr><td><math>x &lt; 2</math></td><td>0/11 (0%)</td><td>NA</td><td>2/4 (50%)</td><td>2/15 (13%)</td></tr><tr><td><math>2 \leq x &lt; 5</math></td><td>13/114 (11%)</td><td>0/1 (0%)</td><td>32/35 (91%)</td><td>45/150 (30%)</td></tr><tr><td><math>5 \leq x &lt; 10</math></td><td>26/71 (37%)</td><td>4/6 (67%)</td><td>15/17 (88%)</td><td>45/94 (48%)</td></tr><tr><td><math>10 \leq x &lt; 20</math></td><td>27/44 (61%)</td><td>NA</td><td>19/19 (100%)</td><td>46/63 (73%)</td></tr><tr><td><math>20 \leq x</math></td><td>5/6 (83%)</td><td>NA</td><td>5/5 (100%)</td><td>10/11 (91%)</td></tr><tr><td>Total</td><td>71/246 (29%)</td><td>4/7 (57%)</td><td>73 (91%)</td><td>148/333 (44%)</td></tr></tbody></table>	SUVmax	1 <sup>st</sup> DP	2 <sup>nd</sup> DP	3 <sup>rd</sup> DP	Total	$x < 2$	0/11 (0%)	NA	2/4 (50%)	2/15 (13%)	$2 \leq x < 5$	13/114 (11%)	0/1 (0%)	32/35 (91%)	45/150 (30%)	$5 \leq x < 10$	26/71 (37%)	4/6 (67%)	15/17 (88%)	45/94 (48%)	$10 \leq x < 20$	27/44 (61%)	NA	19/19 (100%)	46/63 (73%)	$20 \leq x$	5/6 (83%)	NA	5/5 (100%)	10/11 (91%)	Total	71/246 (29%)	4/7 (57%)	73 (91%)	148/333 (44%)
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Background	Methods	Results
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Infos +++  
à extraire des CR  
pour nos études  
rérospectives!

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# Quantification TMTV automatique

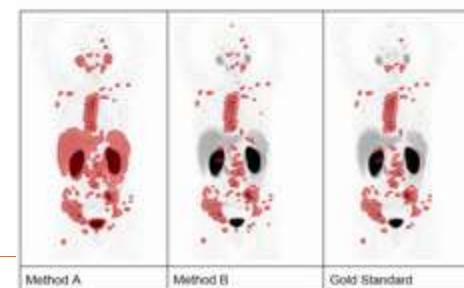
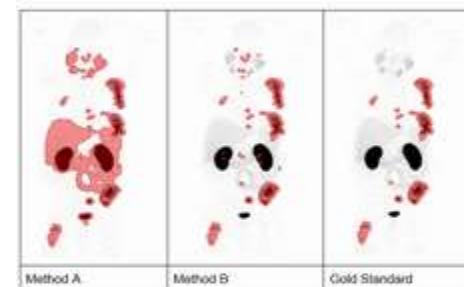
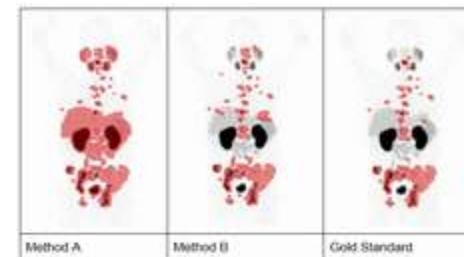
Improved clinical feasibility of total tumor burden quantification on 68Ga-PSMA-11 PET/CT through deep learning auto-segmentation of organs for automatic physiological uptake removal.

Remy Niman, Peter Wilson, Louise Emmett, Sarennya Pathmanandavel and Aaron Nelson

Journal of Nuclear Medicine May 2021, 62 (supplement 1) 1327;

**Objectives:** The utility of total tumor burden (TTB) quantification of 68Ga-PSMA-11 PET/CT for treatment response prediction has been shown (Schmidkonz C et al, 2018; Schmuck S et al, 2017).

The main challenges preventing semi-automatic lesion segmentation from being used in clinical practice are the time required to segment each lesion for patients with high tumor burden and to remove the physiologic uptake captured when using whole body automatic segmentation methods based on SUV thresholds. The objective of this study is to evaluate the use of a deep learning algorithm to create CT-based organ VOIs with a whole body segmentation tool to remove the majority of the PET physiological uptake automatically.



Pour plusieurs traceurs  
(pas que FDG !)  
→ Fixations physiologiques ≠



# Quantification TMTV automatique

**Methods:** 78 68Ga-PSMA-11 PET/CT studies were analyzed from the LuPIN trial dataset. Gold

Standard (GS) TTB VOIs were created using a semi-automatic workflow with an SUV threshold of 3 followed by manual removal of physiological uptake by a credentialed nuclear medicine physician (Crumbaker M et al, 2020). During the GS TTB VOI creation, some regions below 3 SUV were added. Since the liver background activity was typically higher than 3 SUV, a gradient-based method was used to segment liver lesions. 2 whole body segmentation methods were used for evaluation: Method A and Method B. Method A used an SUV threshold of 3 to automatically segment the lesions within the whole body. Lesions of any size were kept within the bone VOI; lesions within the rest of the body were removed if they were less than 0.2ml in size. Method B consisted of lesions from Method A with additional logic applied for removing physiologic uptake. First, a deep learning segmentation method was used to create CT-based organ VOIs for areas of physiological uptake (submandibular glands, parotid glands, bowel, kidneys, bladder) and the liver (NM Cole et al, 2020; A Kruzer et al, 2020; D Mirando, 2020). An atlas-based method was used for the spleen (a physiological uptake VOI) and the pelvic lymph nodes. An HU-based threshold segmentation method was used to create a bone VOI. Areas of increased 68Ga-PSMA-11 activity within the normal organ VOIs were removed automatically. The pelvic lymph node and bone VOIs were used in conjunction with the organ VOIs to prevent activity found in these areas from being removed. Lesions within the liver used a separate SUV threshold of 2 times the SUV mean of an automatically placed 3cm spherical liver reference region.

## Methods

- 78  $^{68}\text{Ga}$ -PSMA-11 PET/CT studies were analyzed from the LuPIN trial dataset<sup>2</sup>.
- Ground Truth segmentation<sup>3</sup>:**
  - Expert nuclear medicine physician used a 3 SUV threshold value and gradient-based segmentation tool (liver lesions only) to segment lesions and manually removed physiological uptake.
- Proposed segmentation:**
  - Automatic whole body PET segmentation method using a 3 SUV threshold value followed by physiological uptake removal using organs segmented based on the CT.

Organ	Type of Uptake	Segmentation Method	Use in Physiological Uptake Removal Algorithm
Submandibular Glands	Physiological	Deep Learning Algorithm	Removed
Parotid Glands	Physiological	Deep Learning Algorithm	Removed
Bowel	Physiological	Deep Learning Algorithm	Removed
Kidneys	Physiological	Deep Learning Algorithm	Removed
Bladder	Physiological	Deep Learning Algorithm	Removed
Spleen	Physiological	Atlas	Removed
Liver	Possible Lesions	Deep Learning Algorithm	Uptake Kept
Pelvic Lymph Nodes	Possible Lesions	Atlas	Uptake Kept
Bone	Possible Lesions	HU-based thresholding	Uptake Kept

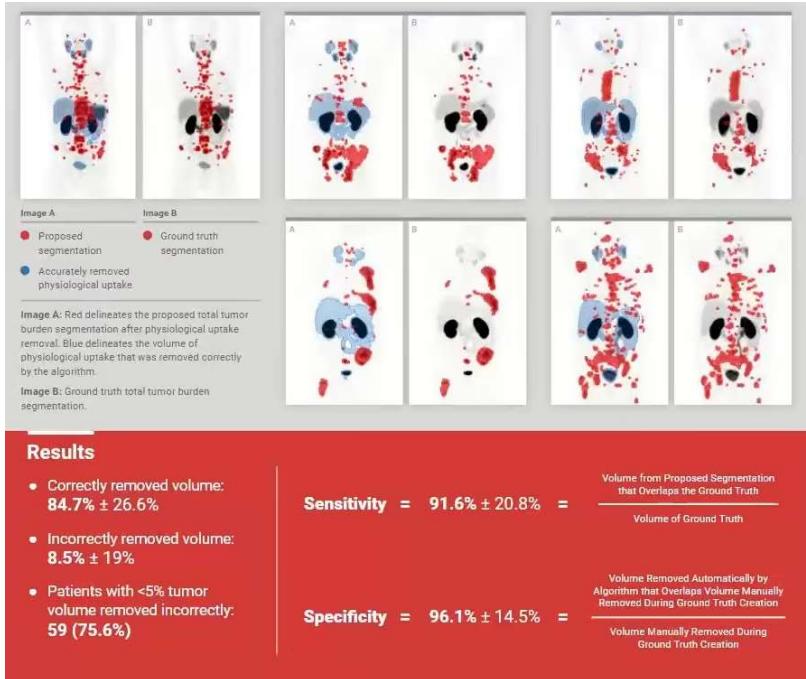


TABLE 1: Various CT organs were segmented in order to remove physiological uptake. Automatically segmented uptake that overlapped the physiological uptake organs was removed automatically. Additional organs (Liver, Pelvic Lymph Nodes, and Bone) were segmented and used to ensure that uptake was not inaccurately removed from those ROIs.

~ “PARS” (=labellisation)  
mais avec des a priori +++



# Quantification TMTV automatique



**Conclusions:** The application of fully automatic organ-based physiological uptake removal results in very similar volumes to those produced by manual editing of TTB volumes. This suggests that minimal additional time would be necessary if used in the clinical workflow. This reduction increases the feasibility of quickly achieving TTB statistics for 68Ga-PSMA-11 PET/CT studies. In future research, we plan to examine the time savings provided by utilizing this automated removal of physiological uptake.

TMTV facile à obtenir → possibilité de tester son utilité

# Données manquantes

## Handling missing values in machine learning to predict major adverse cardiac events: insights from REFINE SPECT registry

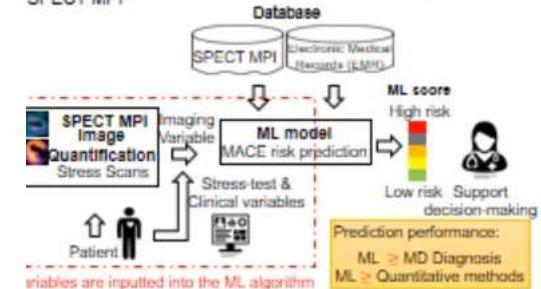
Richard Rios, Robert Miller, Tali Sharir, Andrew Einstein, Mathews Fish, Terrence Ruddy, Philipp Kaufmann, Albert Sinusas, Edward Miller, Timothy Bateman, Sharmila Durbala, Marcelo Di Carli, Joanna Liang, Damini Dey, Daniel Berman and Piotr Slomka

Journal of Nuclear Medicine May 2021, 62 (supplement 1) 1173;

**Objectives:** Machine learning (ML) models, integrating imaging and clinical variables, can improve prediction of major adverse cardiovascular events (MACE) in patients undergoing SPECT MPI. However, in clinical practice, some variables required for ML models may be missing for reasons such as incomplete patient questionnaires or a need to interpret stress data in advance. While algorithms exist for handling missing values, it remains unknown how these methods affect patient-specific risk estimates. We evaluated three methods for handling missing values in ML models for patient-specific prediction of MACE risk.

### Background

Machine learning (ML) models have been used to predict major adverse cardiovascular events (MACE) in patients undergoing SPECT MPI



Some variables with manual entry may be missing for reasons like:

- Incomplete patient questionnaires
- Required interpretation of stress data in advance

**Study aim:**  
To evaluate three methods for handling missing values in ML models for patient-specific prediction of MACE risk

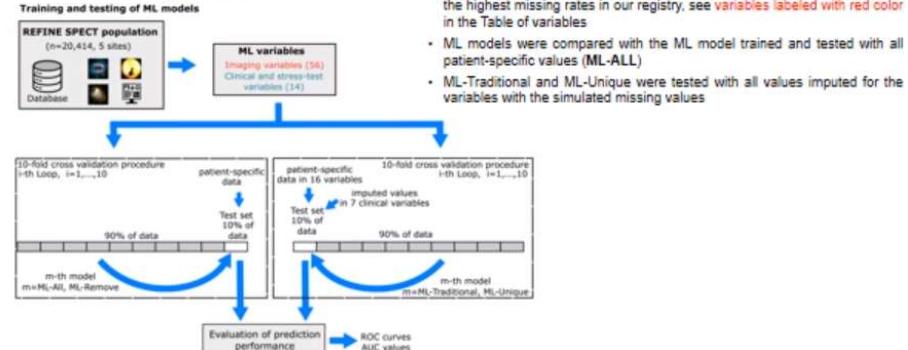
# Données manquantes

**Methods:** This study included 20,414 patients from the multicenter REFINE SPECT registry with images, clinical data, and follow-up for MACE. The average follow-up interval was  $4.7 \pm 1.5$  years. During follow-up, 3,541 patients experienced at least one MACE (3.7% annualized risk): 1,617 deaths, 379 MI, 1,895 revascularizations, and 300 admissions for unstable angina. The dataset included 56 imaging variables and 14 clinical variables. We used extreme gradient boosting (XGBoost) models and evaluated three methods for handling missing values: 1) removal of variables with missing values from the model for MACE prediction (ML-Remove), 2) traditional imputation using the population mean for continuous variables and a distinct missing category for categorical variables (ML-Traditional), and 3) unique category “unknown” imputed for all missing variables, which is the XGBoost’s default method for handling the missing variables (ML-Unique). We compared these ML models to the model built with all variables (ML-All). Training and testing was performed with 10-fold cross validation using the registry data. In each of the testing sets (10% of the registry database), we simulated missing clinical data for the variables which are the most difficult to collect and had the highest missing rates in our registry (*indication for test, resting heart rate, body mass index, ECG response, Symptoms, ST deviation, and clinical response to stress*). ML-Traditional and ML-Unique were tested in a population with all values imputed for those variables. All values were imputed to assess the impact on risk-estimation for an individual patient, where values can either be present or missing. ML-Remove and ML-All were tested with patient specific data. Prediction performance was evaluated using area under the receiver-operating characteristic curve (AUC) and compared with the performance of 4-point scale visual diagnosis and stress total perfusion deficit (TPD) variables.

## Study population

- 20,414 patients from REFINE SPECT registry (Slomka et al. J Nucl Cardiol 2018)
- Clinical data and imaging data collected
- Outcomes was MACE: all cause of mortality, non-fatal myocardial infarction, unstable angina, or late coronary revascularization

## Machine learning methods



We used extreme gradient boosting (XGBoost) and 10-fold cross validation to evaluate three methods for handling missing values:

- Removal of variables with missing values from the model for MACE prediction (ML-Remove)
- Traditional imputation using the population mean for continuous variables and a distinct missing category for categorical variables (ML-Traditional)
- Unique missing category “unknown” imputed for all missing values, which is also XGBoost’s default method (ML-Unique)

List of clinical and stress test variables	
Age	Diabetes Mellitus [0,1]
Stress peak heart rate [beats/min]	ECG response to stress
Previous percutaneous coronary intervention [0,1]	Symptoms [1,2,3,4]
Indication for test [1-23, integer]	Gender
Resting heart rate [bpm]	ST deviation
Pharmacological stress	% of maximum predicted heart rate
Body mass index [ $\text{kg}/\text{m}^2$ ]	Clinical response to stress

3 strategies  
vs  
“all variables”

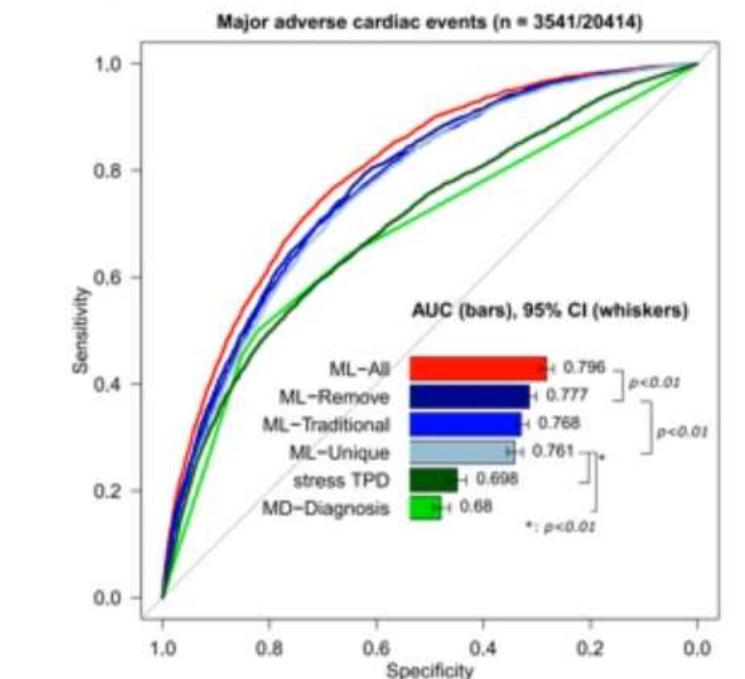
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**Results:** The best MACE risk prediction was obtained when no variables were missing (ML-All AUC: 0.796, 95% confidence interval [CI]: 0.788 to 0.804). The strategy of removing variables during model training obtained the second best prediction performance followed by the traditional imputation method (ML-Remove AUC: 0.777, 95% CI: 0.769 to 0.785 vs ML-Traditional AUC: 0.768, 95% CI: 0.760 to 0.777; p=NS). ML-Unique obtained lower prognostic accuracy, but this was still higher compared with stress TPD and visual diagnosis alone (ML-Unique AUC: 0.761, 95% CI: 0.753 to 0.770 vs stress TPD AUC: 0.698, 95% CI: 0.688 to 0.708 vs. visual diagnosis AUC: 0.680, 95% CI: 0.671 to 0.690, respectively; p < 0.01 for all). **Conclusions:** In clinical practice, missing values reduce the accuracy of ML models when predicting MACE risk. Our results suggest that removing variables with missing values and retraining the model with fewer variables may yield superior accuracy compared to imputation methods.

Conclusions sont transposables sur d'autres jeux de données ? Effectifs plus faibles ?  
Poids des variables manquantes dans le modèle ?

## Results

- 20,414 patients were included
- During mean follow-up of 4.7 years, 3541 patients experienced at least one MACE
- The full ML model (ML-All) had the highest accuracy
- The best method for handling missing values was removal of variables with missing values followed by the traditional imputation method



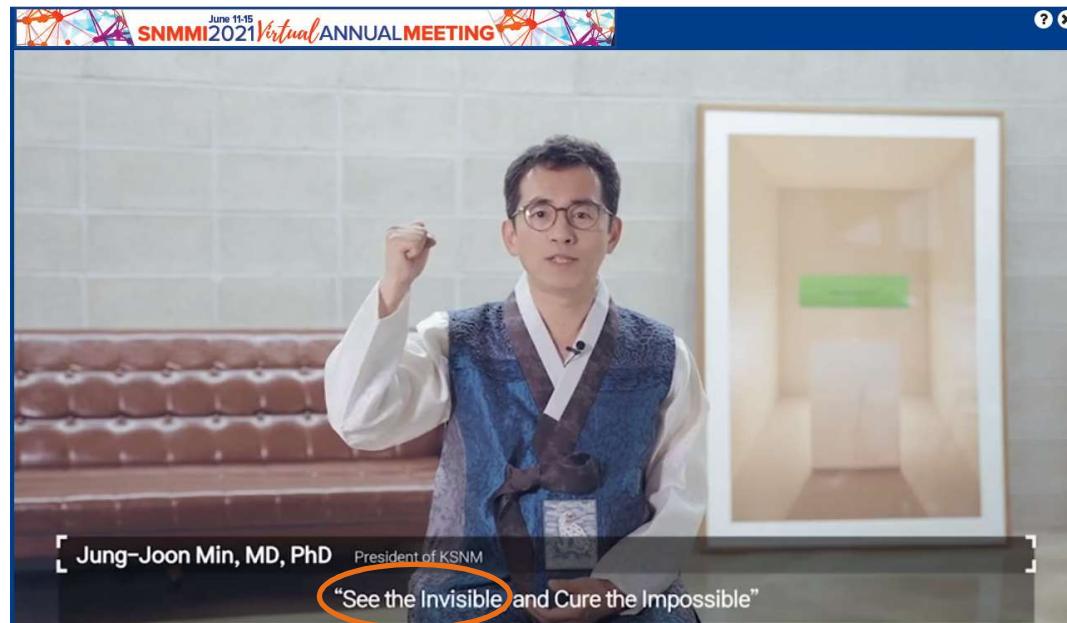
## Pays invite : Corée du Sud



## Pays invite : Corée du Sud



## Pays invite : Corée du Sud





# See the invisible: <sup>18</sup>F-DASA-23

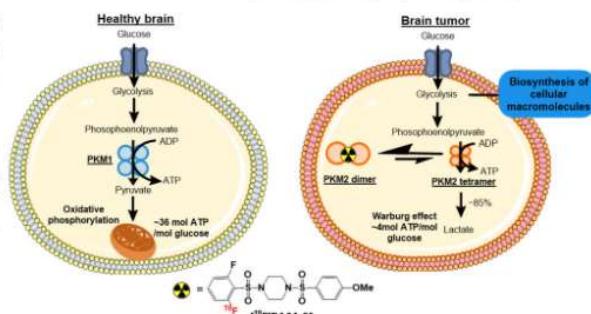
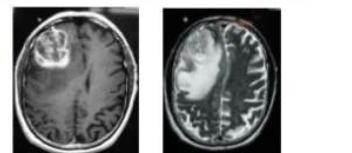
United States  
Published in the United States  
on PubMed

## Initial Clinical Evaluation of <sup>[18]F</sup>DASA-23, a PET Imaging Tracer for Evaluation of Aberrantly Expressed Pyruvate Kinase M2 in Glioblastoma

Corinne Beinat, Chirag Patel, Tom Haywood, Lewis Naya, Jessa Castillo, Bin Shen, Tarik Massoud, Andrei Iagaru, Guido Davidzon, Lawrence Recht and Sanjiv Gambhir  
Journal of Nuclear Medicine May 2021, 62 (supplement 1) 99;

### Molecular Imaging of Brain Tumor Metabolism

- Glioblastoma (GBM) is the most common and lethal primary CNS tumor
- Gold standard for surveillance: brain MRI scans every 2-3 months
- Problem: Brain MRI scans do not discern the cancer's physiologic response to therapy, leading to reactive, rather than pre-emptive, clinical decision making
- Molecular imaging of cancer metabolism provides the opportunity for earlier detection of the tumor's responsiveness to therapy non-invasively



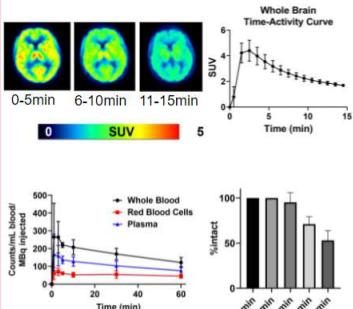
**Introduction:** Pyruvate kinase M2 (PKM2) catalyzes the final step in glycolysis, a key process of cancer metabolism. PKM2 is preferentially expressed by glioblastoma (GBM) cells with minimal expression in healthy brain, making it an important biomarker of cancer glycolytic re-programming. We describe the translation and initial clinical evaluation of a novel positron emission tomography (PET) tracer 1-((2-fluoro-6-[<sup>18</sup>F]fluorophenyl)sulfonyl)-4-((4-methoxyphenyl)sulfonyl)piperazine ([<sup>18</sup>F]DASA-23) in patients with GBM.

**Methods:** [<sup>18</sup>F]DASA-23 was synthesized under current Good Manufacturing Practices United States Food and Drug Administration (FDA) oversight with a molar activity of  $100.47 \pm 29.58$  GBq/ $\mu\text{mol}$  and radiochemical purity of  $>95\%$ . In a phase I trial (NCT03539731), we administered

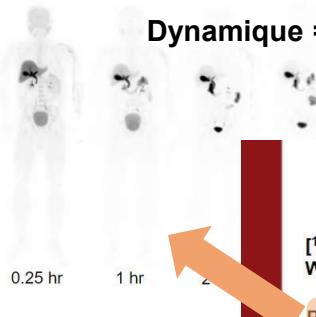


# See the invisible: $[^{18}\text{F}]\text{DASA-23}$

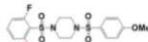
## $[^{18}\text{F}]\text{DASA-23}$ Evaluation in Healthy Volunteers – Part 1a



Dynamique = 3h



## First-In-Human Study of $[^{18}\text{F}]\text{DASA-23}$



$[^{18}\text{F}]\text{DASA-23}$  and PET/MRI Scan in Evaluating Pyruvate Kinase M2 Expression in Patients With Intracranial Tumors or Recurrent Glioblastoma and Healthy Volunteers (NCT03539731)

**Part 1a:** Determine radiation dosimetry and biodistribution in healthy volunteers. Dynamic brain PET/MRI scan over 15 minutes and 4 vertex-to-toe PET/MRI scans over 3 hours

**Part 1b:** Brain uptake and distribution of  $[^{18}\text{F}]\text{DASA-23}$  in healthy volunteers. Dynamic brain PET/MRI scan over 60 minutes

**Part 2:**  $[^{18}\text{F}]\text{DASA-23}$  uptake in intracranial tumor patients. Dynamic brain PET/MRI over 60 minutes

**Part 3:** Response assessment of glioblastoma patients using  $[^{18}\text{F}]\text{DASA-23}$  PET/MRI pre- and post-therapy. Dynamic brain PET/MRI scan over 60 minutes



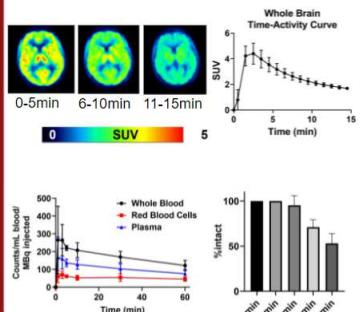
N = 8 sujets sains

# See the invisible: $[^{18}\text{F}]\text{DASA-23}$

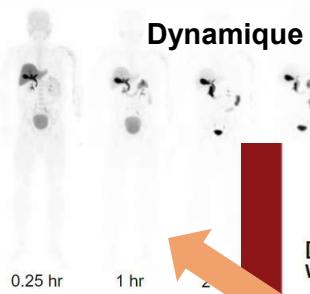
Dynamique = 60 min



## $[^{18}\text{F}]\text{DASA-23}$ Evaluation in Healthy Volunteers – Part 1a



Dynamique = 3h



## First-In-Human Study of $[^{18}\text{F}]\text{DASA-23}$

$[^{18}\text{F}]\text{DASA-23}$  and PET/MRI Scan in Evaluating Pyruvate Kinase M2 Expression in Patients With Intracranial Tumors or Recurrent Glioblastoma and Healthy Volunteers (NCT03539731)

**Part 1a:** Determine radiation dosimetry and biodistribution in healthy volunteers. Dynamic brain PET/MRI scan over 15 minutes and 4 vertex-to-toe PET/MRI scans over 3 hours

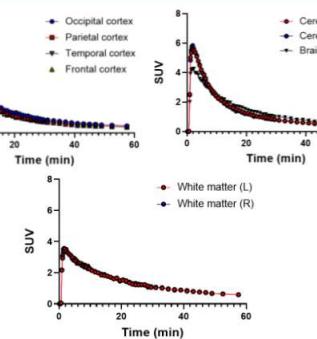
**Part 1b:** Brain uptake and distribution of  $[^{18}\text{F}]\text{DASA-23}$  in healthy volunteers. Dynamic brain PET/MRI scan over 60 minutes

**Part 2:**  $[^{18}\text{F}]\text{DASA-23}$  uptake in intracranial tumor patients. Dynamic brain PET/MRI over 60 minutes

**Part 3:** Response assessment of glioblastoma patients using  $[^{18}\text{F}]\text{DASA-23}$  PET/MRI pre- and post-therapy. Dynamic brain PET/MRI scan over 60 minutes



## $[^{18}\text{F}]\text{DASA-23}$ Brain Distribution – Part 1b



N = 8 sujets sains

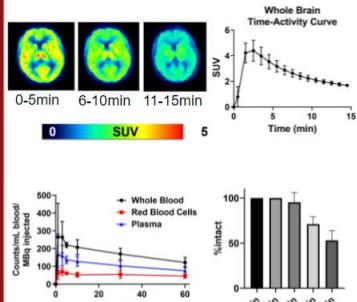


## See the invisible: 18F-DASA-23

## Dynamique = 60 min

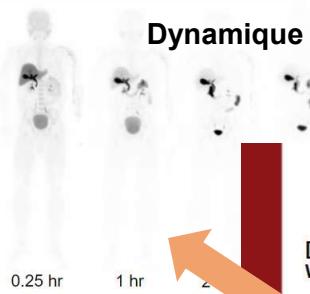


## [<sup>18</sup>F]DASA-23 Evaluation in Healthy Volunteers – Part 1a

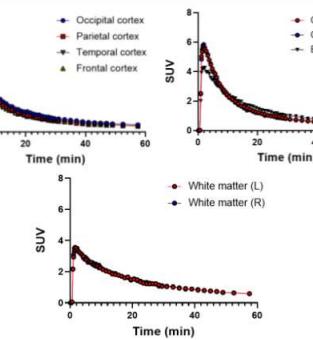
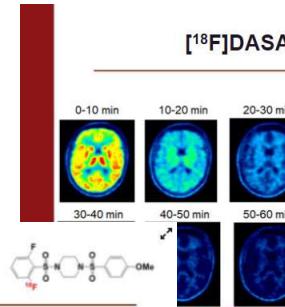
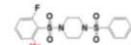


N = 8 sujets sains

## Dynamique = 3h



## First-In-Human Study of [<sup>18</sup>F]DASA-23



**Part 1a:** Determine radiation dosimetry and biodistribution in healthy volunteers. Dynamic brain PET/MRI scan 15 minutes and 4 months after PET/MRI scan 2 hours

**Part 1a:** Determine radiation dosimetry and biodistribution in healthy volunteers. Dynamic brain PET/MRI scan (15 minutes) and 4 months later PET/MRI scan (2 hours).

#### **Part 1b: Brain uptake and distribution of [<sup>18</sup>F]DASA 23 in healthy volunteers. Dyn**

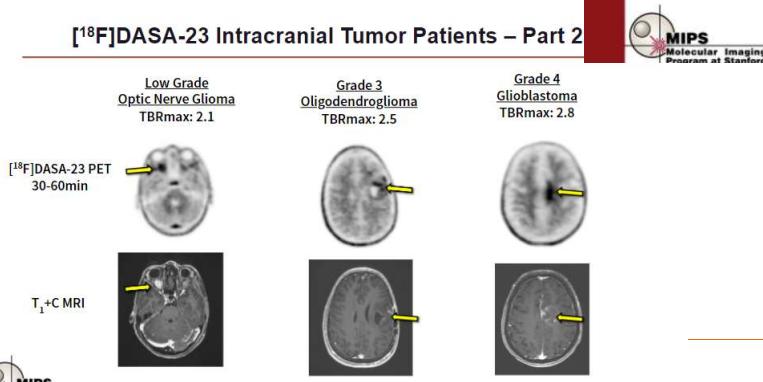
## Part 1b: Brain uptake and distribution of [<sup>18</sup>F]DASA-23 in healthy volunteers. Dynamic brain

PET/MRI scan over 60 minutes

## Part 2: [<sup>18</sup>F]DASA-23 uptake in intracranial tumor patients. Dynamic brain PET/MRI over 60 minutes

### Part 3: Response assessment of glioblastoma patients using [<sup>18</sup>F]DASA-23 PET/MRI pre- and post-therapy. Dynamic brain PET/MRI scan over 60 minutes

## [<sup>18</sup>F]DASA-23 Intracranial Tumor Patients – Part 2

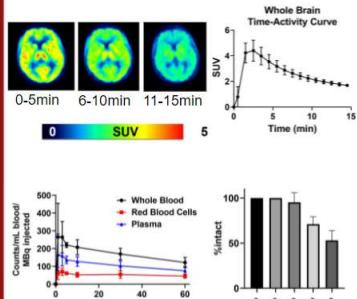


## See the invisible: 18F-DASA-23

## Dynamique = 60 min

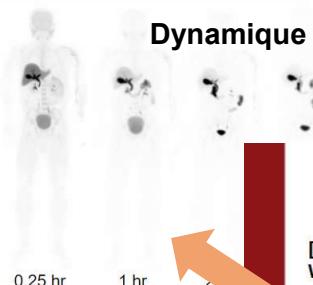


## [<sup>18</sup>F]DASA-23 Evaluation in Healthy Volunteers – Part 1a

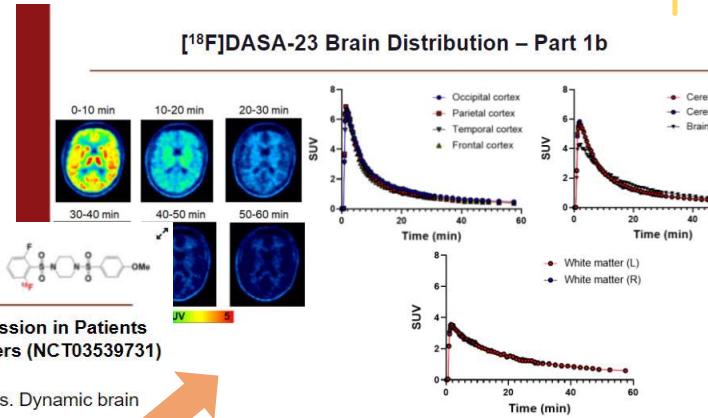


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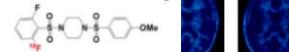
## Dynamique = 3h



[<sup>18</sup>F]DASA-23 Brain Distribution – Part 1b



## First-In-Human Study of [<sup>18</sup>F]DASA-23



**[<sup>18</sup>F]DASA-23 and PET/MRI Scan in Evaluating Pyruvate Kinase M2 Expression in Patients With Intracranial Tumors or Recurrent Glioblastoma and Healthy Volunteers (NCT03539731)**

**Part 1a:** Determine radiation dosimetry and biodistribution in healthy volunteers. Dynamic brain

PET/MRI scan over 15 minutes and 4 vertex-to-toe PET/MRI scans over 3 hours

**Part 1b:** Brain uptake and distribution of [<sup>18</sup>F]DASA-23 in healthy volunteers. Dynamic brain

**PET/**MRI scan over 60 minutes  
**Part 2:**  $[^{18}\text{F}]$ DASA-23 uptake in intracranial tumor patients. Dynamic brain PET/MRI over 60 minutes

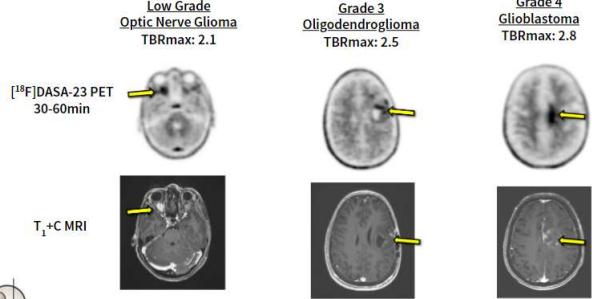
**Part 3:** Response assessment of glioblastoma patients using [ $^{18}\text{F}$ ]DASA-23 PET/MRI pre- and post-therapy. Dynamic brain PET/MRI scan over 60 minutes

eriod (Fig. 1b). In patients with intracranial malignancies,  $[^{18}\text{F}]$ DASA-23 successfully visualized contrast-enhancing tumors with a trend towards increasing uptake with increasing tumor grade (Fig. c). In a GBM patient,  $[^{18}\text{F}]$ DASA-23 SUVmean and SUVmax were 2.0 and 2.2, respectively. The absence of PKM2 within healthy brain resulted in a high tumor-to-brain ratio (TBR) of 2.5 and a

## [<sup>18</sup>F]DASA-23 Intracranial Tumor Patients – Part 2



[<sup>18</sup>F]FDG PET/CT Intracranial Tumors



[<sup>18</sup>F]DASA-23 Response Assessment – Part 3

The figure displays six panels arranged in a 2x3 grid. The top row shows **T<sub>1</sub>+C MRI**, **[<sup>18</sup>F]FDG PET/CT**, and **[<sup>18</sup>F]DASA-23 PET**. The bottom row shows **Pre-therapy**, **Post-therapy**, and **3 month follow-up**. A color scale bar at the bottom indicates **0** to **500 SUV**. The **[<sup>18</sup>F]DASA-23 PET** images show a yellow arrow pointing to a region of high uptake and a red arrow pointing to a region of low uptake.

Panel	Image Type	Time Point
1	T <sub>1</sub> +C MRI	15 days prior to treatment
2	[ <sup>18</sup> F]FDG PET/CT	12 days prior to treatment
3	[ <sup>18</sup> F]DASA-23 PET	3 days prior to treatment
4	[ <sup>18</sup> F]DASA-23 PET	6 days post treatment
5	T <sub>1</sub> +C MRI	6 days post treatment
6	T <sub>1</sub> +C MRI	90 days post treatment

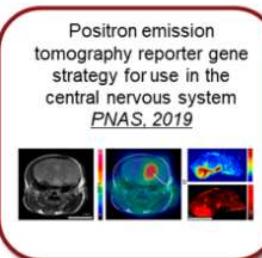
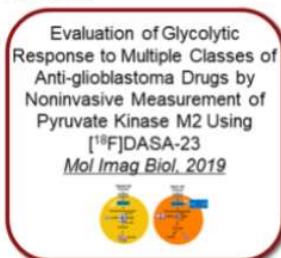
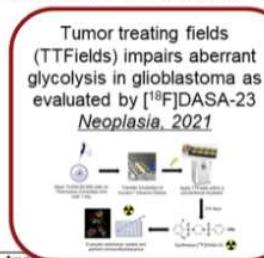
Day 0: TMZ and Avastin



# See the invisible: 18F-DASA-23

## Conclusions and Ongoing Work

- $[^{18}\text{F}]$ DASA-23 is a new radiotracer for PKM2 and is approved for evaluation in healthy volunteers and patients
- $[^{18}\text{F}]$ DASA-23 is safe to use in humans with no adverse events or changes in blood labs
- We are recruiting patients to evaluate the ability of  $[^{18}\text{F}]$ DASA-23 to detect early therapeutic response in newly diagnosed and recurrent glioblastoma (NCT03539731)
- Other studies involving this tracer include:



### PRIMARY OBJECTIVES:

- I. Determine whether the fluorine F 18 DASA-23 ( $[^{18}\text{F}]$ DASA-23) PET scan signal change from pre-therapy to one week after initiation of therapy can predict the tumor's responsiveness to therapy and 6 month progression free survival (PFS6), in suspected recurrent glioblastoma.

Study Type [I](#) : Interventional (Clinical Trial)

Estimated Enrollment [I](#) : 25 participants

Allocation: Non-Randomized

Intervention Model: Sequential Assignment

Masking: None (Open Label)

Primary Purpose: Diagnostic

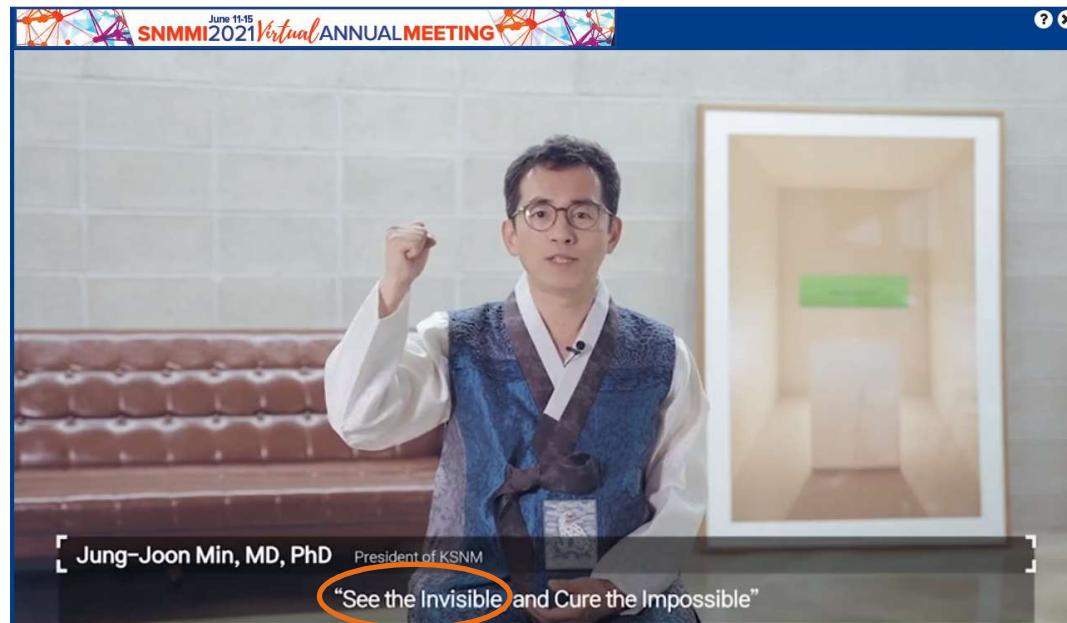
Official Title: A Phase I Study of  $[^{18}\text{F}]$ DASA-23 as a PET Tracer for Evaluating Pyruvate Kinase M2 (PKM2) Expression in Healthy Volunteers and in Patients With Intracranial Tumors

Actual Study Start Date [I](#) : April 23, 2018

Estimated Primary Completion Date [I](#) : July 30, 2022

Estimated Study Completion Date [I](#) : December 30, 2022

## Pays invite : Corée du Sud



# See the invisible: FAPI+++

## More on FAPI

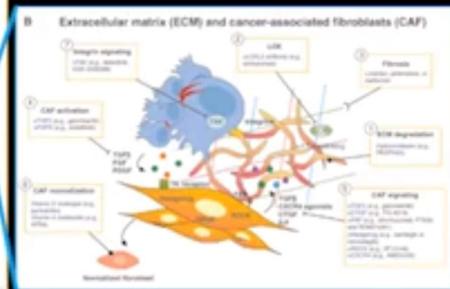
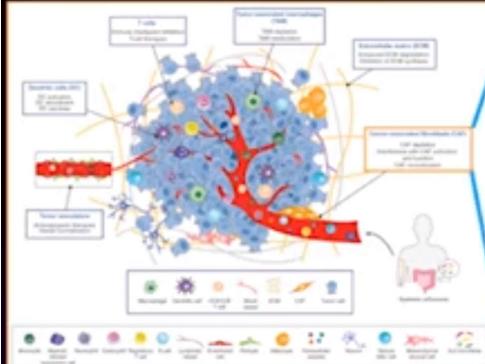


SNMMI 2019 – Image of the Year  
Krebsen, Hoberman, Glaser et al.

### Meanwhile

- Numerous case reports and small clinical studies
  - Utility for diagnosis, staging, RT planning, change in pt. management
  - Lung, pancreas, gastric, lower GI, HNC incl. ACC and NPC, sarcoma, peritoneal carcinomatosis
- Non-cancer
  - ECD, thyroiditis, benign pancreas lesions, pulmonary fibrosis, FD, solitary fibrous tumor, elastofibroma...
  - Post-MI fibroblast activation

## Tumor Microenvironment – CAFs and FAP



Rejarto et al. *Cancer Disc* 2021;11:83-88

## CAFs and FAP

### Cancer Associated Fibroblasts

- Subpopulation of activated fibroblasts in the TME
- Relevant for
  - Angiogenesis
  - Cell migration and metastasis
  - Immune suppression
  - Resistance to chemotherapy, anti-androgens
- Actions
  - Secretion of growth factors, chemokines
  - Remodelling of ECM
  - Cooperative metabolism w/cancer cells

### Fibroblast Activating Protein

- Only expressed on activated fibroblasts
- Transmembrane glycoprotein
- Enzymatic activity as peptidase
- Various physiologic roles in normal development
- Prognostic marker in cancer
- Can be targeted with AB, AB-drug-conjugates, immunocytokines, CAR T-cells, activatable prodrugs, enzyme inhibitors...
  - Radiolabeled FAPI for theranostics
  - Poss. synergistic with immunotherapies

FAP = *Fibroblast Activation Protein*  
Très exprimée dans le stroma tumoral

institut  
**Curie**



## See the invisible: FAPi+++

### Validation of FAPi PET biodistribution by immunohistochemistry in patients with solid cancers: a prospective exploratory imaging study

Christine Mona, Matthias Benz, Firas Hikmat, Tristan Grogan, Katharina Lueckerath, Ali Aria Razmaria, Rana Riahi, Roger Slavik, Mark Grigis, Giuseppe Carlucci, Kimberly Kelly, Johannes Czernin, David Dawson and Jeremie Calais

Journal of Nuclear Medicine May 2021, 62 (supplement 1) 1000;

**Introduction:** Fibroblast activation protein (FAP)-expressing cancer associated fibroblasts (CAFs), a major component of tumor stroma, confer treatment resistance, promote local progression, metastasis and immunosuppression. Because FAP is selectively expressed in the tumor stroma of many cancers, radiolabeled small molecule ligands targeting FAP are being explored for their use as pan-cancer theranostic agents. The objective was to define the incidence and degree of FAP expression by immunohistochemistry (IHC) across various cancers using Tissue Microarrays (TMAs) and to explore whether gallium-68 FAPi-46 PET image biodistribution faithfully reflects tumor FAP expression from resected tumor and non-tumor specimens.

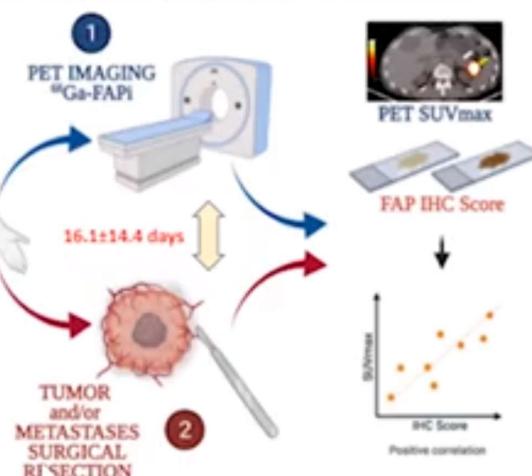
# See the invisible: FAPI+++

**Methods:** This study was a prospective, exploratory, open-label, single-center imaging trial in cancer patients conducted in 2020. Referred volunteer patients scheduled to undergo surgical resection of the primary tumor and/or metastases were eligible. Patients underwent one whole body <sup>68</sup>Ga-FAPi-46 PET/CT scan. Subsequently, patients underwent surgical resection of the primary tumor and/or metastasis. The **Outcome Measure was the Correlation of <sup>68</sup>Ga-FAPi-46 PET maximum standardized uptake value (SUVmax) with FAP IHC score in cancer and non-cancer tissue.** **Trial Registration:** ClinicalTrials.gov Identifier: NCT04147494.

## <sup>68</sup>Ga-FAPi-46 PET biodistribution by immunohistochemistry in patients with solid cancers

Mona et al.  
Ahmanson Translational Theranostics Division  
Department of Molecular and Medical  
Pharmacology  
David Geffen School of Medicine  
University of California - Los Angeles

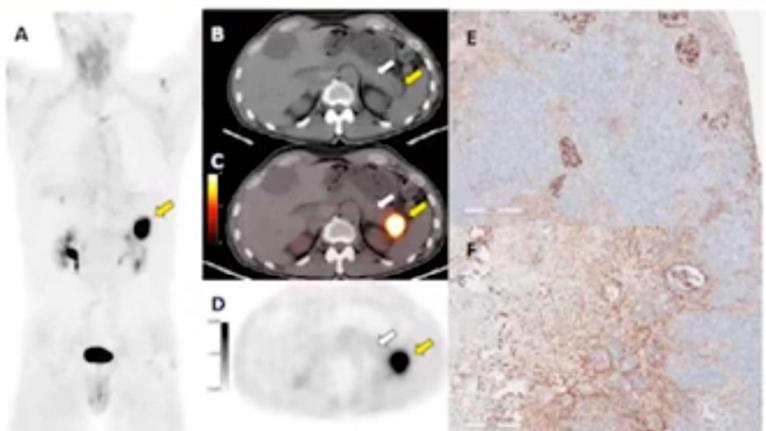
	n=
<b>Total</b>	<b>15</b>
Colorectal	4
Head and Neck	3
Pancreas	2
Breast	2
Esophagus	2
Stomach	1
Uterus	1



UCLA

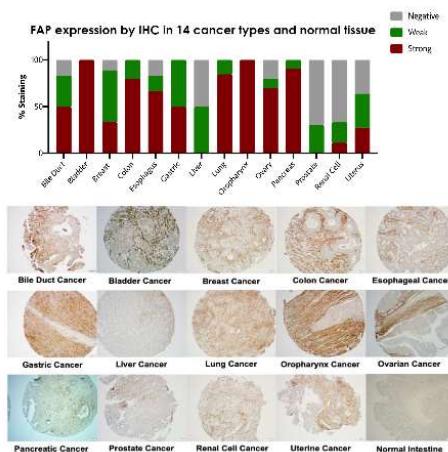
Prospective exploratory trial NCT04147494

Abstr. 1000



A 65-year-old male patient with pancreatic ductal adenocarcinoma. pancreatic tail ductal adenocarcinoma lesion, yellow arrows (B: transaxial CT, (C and D: transaxial PET/CT and PET, SUV<sub>max</sub> 15.69 and SUV<sub>mean</sub> 12.51, respectively). FAP IHC on representative histologic sections demonstrated variable negative to weak FAP expression in normal pancreatic parenchyma with a subpopulation of cells in normal islets consistently showing strong FAP expression (E). Moderate to strong FAP expression was noted for tumor tissue (F). White arrows depict resected normal pancreas region.

Abstr. 1000

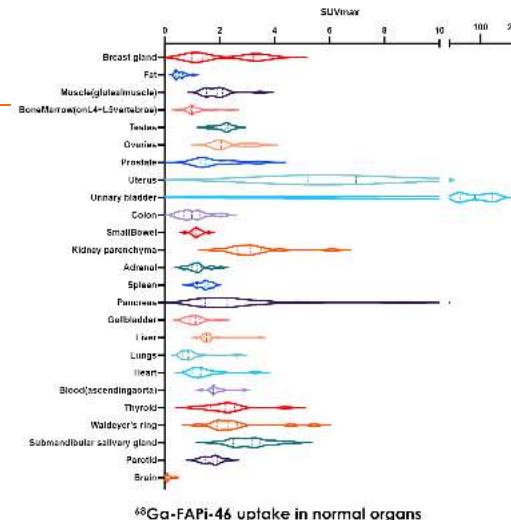


# See the invisible: FAPI+++

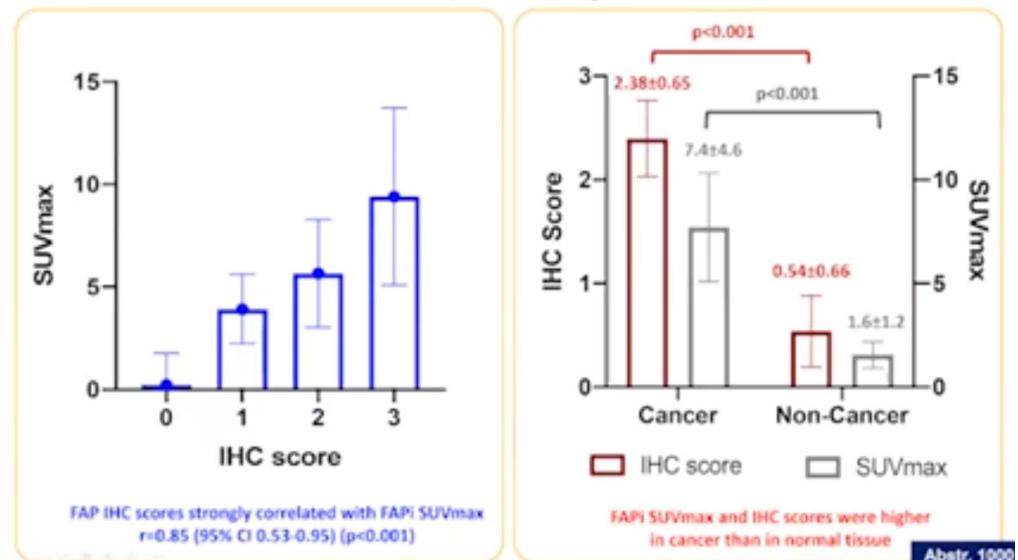


**Results:** The incidence of FAP expression in TMAs from 14 cancers ranged from 25 to 100% (mean  $76.6 \pm 25.3$ ). 15 patients with the following cancer types were included: colorectal (n=4), head and neck (n=3), pancreas (n=2), breast (n=2), stomach (n=1), esophagus (n=2) and uterus (n=1). All 15 patients subsequently underwent surgery after the scan with a mean time interval of  $16.1 \pm 14.4$  days (range 1 - 50 days). Tumor resection was not attempted in 2 patients because unresectable. FAPI SUVmax and IHC score were higher in cancer tissue than in normal tissue: mean FAPI SUVmax  $7.4 \pm 4.6$  (range 1.5-15.9) vs  $1.6 \pm 1.2$  (range 0.4-5.1), ( $p < 0.001$ ) and mean FAP IHC score  $2.38 \pm 0.65$  vs  $0.54 \pm 0.66$  ( $p < 0.001$ ), respectively. The FAP IHC score was positively correlated with FAPI SUVmax (pairwise  $p = 0.001$ , repeated measures correlations  $r = 0.85$  (95% CI 0.53-0.95),  $p < 0.001$ ).

15 patients/7 types de cancer

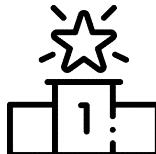


$^{68}\text{Ga}$ -FAPI-46 uptake in normal organs





# See the invisible: FAPI+++



**Conclusions:** The  $^{68}\text{Ga}$ -FAPI-46 PET biodistribution across multiple cancers reflects FAP expression as determined by IHC. This translational validation paves the way for large scale prospective trials on the use of  $^{68}\text{Ga}$ -FAPI-46 PET/CT as a biomarker and stratification tool for FAP-targeted therapies.

## CONCLUSION:

- FAPI uptake tends to be **significantly higher in cancer** tissues vs. non-cancer tissues
- $^{68}\text{Ga}$ -FAPI-46 PET biodistribution across multiple cancers **strongly correlates with fibroblast activation protein tissue expression** as measured by immunohistochemistry
- Staining intensity was higher in stromal areas within and immediately adjacent to the malignant epithelial compartment of tumors.
- FAP staining was exclusively **confined to the tumor-associated stromal compartment** in most patients
- The highest FAP IHC scores were observed in pancreatic, esophageal and breast cancer.

## Additional FAPI Studies

**Abstr. 20: peritoneal carcinomatosis**

- N=46; SOR: surgery, biopsy or FU
- FAPI+ lesions:
  - FAPI uptake > FDG uptake
  - Size < FDG+ lesion
  - Particular advantage in gastric and colon cancer lesions

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Role of  $^{68}\text{Ga}$ Ga-201A-FAP-46 PET/CT in the evaluation of peritoneal carcinomatosis and comparison with  $^{18}\text{F}$ FDG PET/CT

**Abstr. 125: gastric, duodenal and colorectal cancers**

- N=35
- Higher uptake, better sens. than FDG for primaries and metastases

[View Abstract](#)

[View Article](#)

Comparison of  $^{68}\text{Ga}$ -FAPI and  $^{18}\text{F}$ -FDG Uptake in Gastric, Duodenal, and Colorectal Cancers

**Abstr. 124: nasopharynx cancer**

- N=15
- Better detection of skull base invasion
- May detect metastases outside H&N → upstaging

[View Abstract](#)

[View Article](#)

A head-neck comparison of  $^{68}\text{Ga}$ EDTA-FAP-46 and  $^{18}\text{F}$ -FDG PET/CT in patients with nasopharyngeal carcinoma: a prospective study

**Abstr. 1086: nasopharynx cancer; FDG-FAPI-MRI**

- N=45
- Uptake FAPI > FDG in primary, nodes, metastases
- Upstaging in 26% of pts; & management in 15%
- Some incremental benefit to MRI (skull base)
- Post-Rx: fewer FP than FDG

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[View Article](#)

Clinical utility of  $^{68}\text{Ga}$ Ga-labeled fibroblast activation protein (FAP)-positive emission tomography/positron tomography for primary staging and recurrence detection in nasopharyngeal carcinoma

## Pays invite : Corée du Sud





# Cure the impossible: TheraP

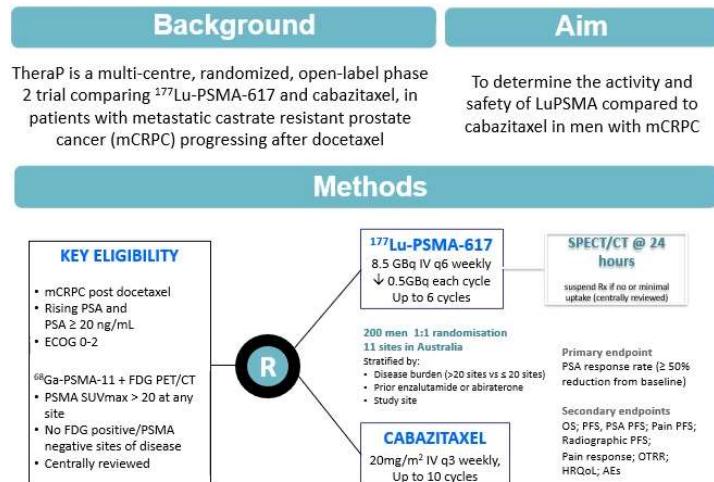
## $^{177}\text{Lu}$ -PSMA-617 versus Cabazitaxel in Metastatic Castration-Resistant Prostate Cancer: a randomised, open-label, phase 2 trial (TheraP)

Michael Hofman, Louise Emmett, Amir Iravani, Shahneen Sandhu, Anthony Joshua, David Pattison, Jeffrey Goh, Ian Kirkwood, Thean Hsiang Tan, Roslyn Francis, Siobhan Ng, Natalie Rutherford, Craig Gedye, Andrew Scott, Sze-Ting Lee, Andrew Weickhardt, Shakher Ramdave, Edmond Kwan, Arun Azad, William Macdonald, Andrew Redfern, Alison Zhang, Martin Stockler, Andrew Martin and Ian Davis

Journal of Nuclear Medicine May 2021, 62 (supplement 1) 1703;

**Objectives:**  $^{177}\text{Lu}$ -PSMA-617 is a radiolabelled small molecule that delivers  $\beta$ -radiation to cells expressing prostate specific membrane antigen (PSMA), with promising activity and safety in metastatic castration-resistant prostate cancer (mCRPC). We compared  $^{177}\text{Lu}$ -PSMA-617 and cabazitaxel in a randomised phase 2 trial.

**Methods:** Men with mCRPC progressing after docetaxel with high PSMA tumour-expression were randomized to  $^{177}\text{Lu}$ -PSMA-617 (6-8.5 GBq intravenously 6 weekly up to 6 cycles) vs cabazitaxel (20 mg/m<sup>2</sup> intravenously 3 weekly up to 10 cycles) at 11-sites in Australia. The primary endpoint was prostate specific antigen (PSA) response rate defined by  $\geq 50\%$  reduction (PSA50-RR). Secondary endpoints included progression-free survival (PFS) (PSA and radiographic PFS), objective response rate (ORR) (RECIST 1.1), adverse events (CTCAE v4.03), patient-reported outcomes (PROs) (EORTC QLQ-C30, PDF), and overall survival (OS). This trial is registered in ClinicalTrials.gov, NCT03392428.

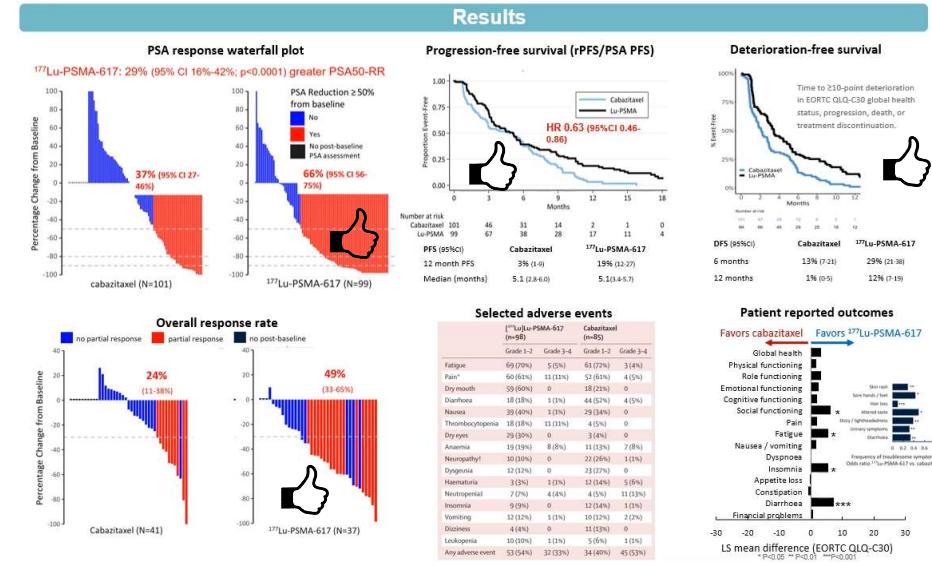
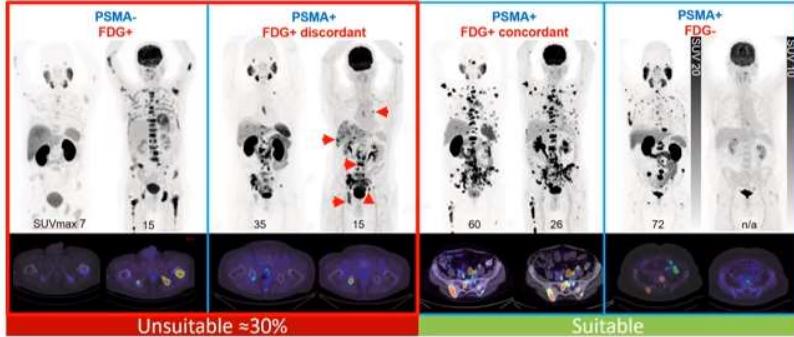




# Cure the impossible: TheraP

**Results:** 200 of 291 men identified as eligible on PET imaging were randomised to  $^{177}\text{Lu}$ -PSMA-617 (N=99) or cabazitaxel (N=101). 91% had received prior androgen receptor-directed therapy (ARDT). PSA50-RR was significantly higher in those assigned  $^{177}\text{Lu}$ -PSMA-617 versus cabazitaxel (66% [95%CI,56-75%] vs 37% [95%CI,27-46%];  $P<0.001$ ). PFS was significantly longer in those assigned  $^{177}\text{Lu}$ -PSMA-617 than cabazitaxel (rates at 1yr 19% [95%CI,12-27%] vs 3% [95%CI,1-9%], hazard ratio (HR) 0.63 [95%CI,0.46-0.86;  $P=0.003$ ). ORR in 78 men with measurable disease was significantly higher with  $^{177}\text{Lu}$ -PSMA-617 than cabazitaxel (49% vs 24%, RR 2.12;  $P=0.026$ ). Follow-up remains immature for OS. Grade 3-4 adverse events occurred in 32/98 (33%) with  $^{177}\text{Lu}$ -PSMA-617 vs 45/85 (55%) with cabazitaxel. Overall quality-of-life and health status were similar, with significantly better outcomes for multiple PRO domains including diarrhoea, fatigue, social functioning, insomnia, hair loss, skin rash and sore hands/feet with  $^{177}\text{Lu}$ -PSMA-617.

## Patient screening

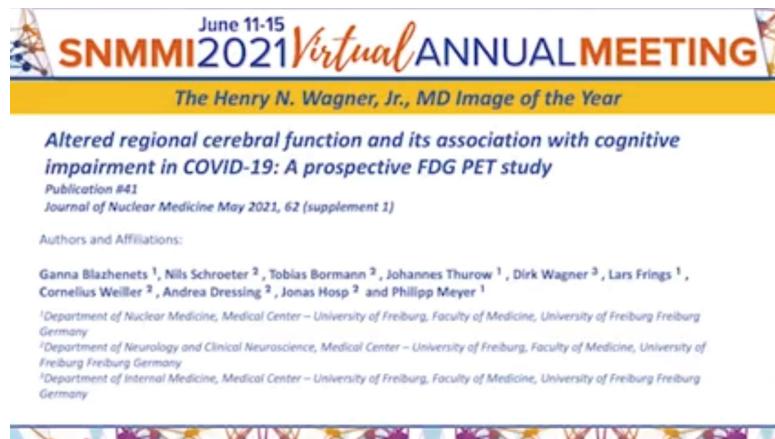


**Conclusions:**  $^{177}\text{Lu}$ -PSMA-617 compared to cabazitaxel in men with mCRPC led to significantly higher PSA and ORR response rates, longer PFS, less grade 3 or 4 adverse events and significant improvements in several PRO domains.

# Image de l'année

## Altered regional cerebral function and its association with cognitive impairment in COVID-19: A prospective FDG PET study

Ganna Blazhenets, Nils Schroeter, Tobias Bormann, Johannes Thurow, Dirk Wagner, Lars Frings, Cornelius Weiller, Andrea Dressing, Jonas Hosp and Philipp Meyer  
Journal of Nuclear Medicine May 2021, 62 (supplement 1) 41;



**Conclusions:** Neocortical dysfunction accompanied by cognitive impairment was detected in two-thirds of inpatients with subacute COVID-19. Although a significant recovery of regional neuronal function and cognition can be clearly stated, residuals are still measurable in some patients six month after manifestation of COVID-19. In consequence, post-COVID-19 patients with persistent cognitive complaints should be presented to a neurologist and possibly allocated to cognitive rehabilitation programs.

