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Non-Invasive Determination of Blood Input Function to Compute Rate of Myocardial Glucose Uptake from Dynamic FDG PET Images of Rat Heart In Vivo: Comparative Study Between the Inferior Vena Cava and the Left Ventricular Blood Pool with Spill Over and Partial Volume Corrections

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Abstract

The purpose of this work was to compute blood input function from the inferior vena cava (IVC) with partial volume (PV) corrections and compare to that obtained from the left ventricular blood pool (LVBP) with spill-over (SP) and partial volume (PV) corrections. These were then used to compute and validate rates of myocardial 2-deoxy-2-[18F]fluoro-D-glucose (FDG) uptake (Ki) from dynamic positron emission tomography (PET) images of rat hearts *in vivo* in comparison to that obtained from invasive arterial blood sampling. Whole body 60 minute dynamic FDG PET/CT imaging of n=8 control Wistar Kyoto (WKY) rats were performed using Albira trimodal PET/CT/SPECT scanner. Image derived blood input function (IDIF) obtained from IVC corrected for PV averaging (IVC-PV) and IDIF from the left ventricular blood pool (LVBP) with SP and PV corrections (LVBP-SP-PV) were computed. Next, computed Ki (indirect comparison) in a 5-parameter (using IVC-PV) and a 15-parameter (using LVBP-SP-PV) 3-compartment models in WKY rat hearts *in vivo* were compared to that obtained using arterial blood sampling reported in literature in control Sprague Dawley (SD) rats. Using IVC-PV in a 3-compartment 5-parameter model resulted in a ~46% deviation in the mean computed Ki compared to that obtained with LVBP-SP-PV in a 3-compartment 15-parameter model with a ~57% deviation in the mean computed Ki. The mean computed Ki in WKY rat hearts using the above methods, however, did not differ significantly to that obtained from invasive arterial blood sampling in SD rat hearts (p=0.91 for IVC-PV and p=0.58 for LVBP-SP-PV). Hence, Ki obtained in WKY rat hearts with input curve from IVC (IVC-PV) in a dynamic FDG PET scan is comparatively more repetitive to that obtained from the LVBP (LVBP-SP-PV). Ki computed using both the methods, however, agree well with each other and that obtained using arterial blood sampling.

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Conflict of Interest

The authors declare that they have no conflict of interest.

Keywords

Dynamic FDG PET; Blood input function; Inferior vena cava; Left ventricular blood pool; Spill over and Partial Volume corrections

1. Introduction

Determination of blood input function non-invasively from dynamic 2-[18F] fluoro-2-deoxy-D-glucose positron emission tomography (FDG PET) images of rodent hearts has been challenging due to small size of the left ventricle and the limited spatial resolution of small animal scanners. This results in incomplete radioactivity recovery or partial volume (PV) averaging for both the myocardium and the blood pool resulting in significant spill-out contamination from the blood to the tissue at the early time points and spill-in from the tissue to the blood at the late time points in a dynamic FDG PET scan. Non-invasive estimation of the blood input function thereby becomes challenging due to multi-parameter nature of the optimization problem in a 3-compartment model (Krivokapich *J et al* 1982). Prior work by others (Fang and Muzic 2008) and from our laboratory (Zhong and Kundu 2013) developed a 15-parameter model for simultaneous estimation of model corrected blood input function (MCIF) and compartment model parameters to determine rate of myocardial FDG uptake, K_i . This method, however, utilized a deterministic method (Interior Reflective Newton (IRN)) for optimization of the objective function, which determines only a local minimum. This optimization technique is sensitive to changes in the bounds and initial guess values of the kinetic parameters. Recent work from our laboratory (Li and Kundu 2018) developed a hybrid method combining stochastic (Artificial Immune System (AIS)) and deterministic methods (IRN) for non-invasive estimation of K_i . This method is more robust in that it determines a global minimum, thereby making it less sensitive to changes in the bounds and guess values of the kinetic parameters. However, both the above described methods developed in rodents involve optimization of 15 parameters, which may be challenging in a clinical setting.

The goal of this study was to obtain image derived blood input function and downstream K_i , from the inferior vena cava (IVC) and compare to that obtained from the left ventricular blood pool (LVBP) of dynamic FDG PET images of control Wistar Kyoto (WKY) rats *in vivo* with spill-over (SP) and PV corrections.

2. Materials and Methods

Eight (n=8) control male WKY rats, purchased from Charles River (Kingston, NY), were housed under controlled conditions (temperature $21\pm 1^\circ\text{C}$, humidity $60\pm 10\%$, 12-hours light/12-hours dark cycle, and free access to standard rat chow and water). Animal experiments were approved by the Institutional Animal Care & Use Committee of the University of Virginia.

2.1 Dynamic FDG PET Imaging:

Eight male WKY rats, 2 months of age, were imaged using a state-of-art Albira trimodal PET/SPECT/CT scanner (González *et al* 2016). The Albira PET imager is a 3-ring scanner with an axial field of view (FOV) of 150 mm and trans-axial FOV of 80 mm, thereby enabling whole body PET imaging of 2-month old rats weighing around 200 g. The homogenous intrinsic spatial resolution of the scanner was reported to be better than 1.2 mm in the whole FOV of 80 mm. Dynamic FDG PET imaging of the rats was performed using a similar protocol as described earlier in mice (Zhong and Kundu 2013, Zhong *et al* 2013) and more recently in rats (Li *et al* JAHA 2019). Briefly, a 60 minute list mode (LM) acquisition, under 2% isoflurane anesthesia with a respiratory pillow attached to the chest, was initiated followed by intravenous injection of ~200–300 uCi of FDG via a tail-vein catheter over a period of ~10–20 seconds. The rats were fasted for 5–6 hours prior to PET imaging. A 3-bed CT scan followed the emission scan for attenuation correction. A small-animal gating and monitoring system (Small Animal Instruments, Inc., model 1025L for PET) was used for continuously monitoring respiration and core body temperature (using a rectal probe). The LM data was then sorted into 23 time bins (frames, time (s): 11,8; 1,12; 2,60;1,180;8,400) and the sinograms reconstructed with attenuation correction using Maximum Likelihood Expectation Maximization (MLEM) algorithm with 6 iterations and 0.75 mm isotropic voxel resolution (Huang *et al* 2018).

2.2 Input curve from the Inferior Vena Cave (IVC) and Partial volume corrections (IVC-PV):

Regions of interest (ROI) were drawn in the early time frame in the region corresponding to the inferior vena cava (IVC) (median volume: 4.5 mm³) and time activity curve (TAC) generated (blood TAC).

2.3 Partial volume corrections:

Partial volume (PV) corrections was performed on the blood TAC using a recovery coefficient (RC) that was generated for each rat by performing a convolution of the finite size of the IVC with a Gaussian distribution representing the spatial resolution of the Albira PET system. Since PV averaging is a function of geometry and does not depend on time, the blood TAC for each rat was then boosted by its corresponding RC.

2.4 Kinetic modeling:

The blood TAC obtained from the IVC corrected for PV averaging was used in a 3-compartment model (Zhong and Kundu 2013) to compute rate of myocardial FDG uptake, K_i . The formalism for the 3-compartment model was written as follows:

$$C_m(t) = \frac{1}{(t_2 - t_1)} \cdot [(1 - TBV)\rho \cdot \int_{t_1}^{t_2} C_T(t)dt + TBV \int_{t_1}^{t_2} C_p(t)dt] \quad (1)$$

where, $C_m(t)$ is the activity concentration over time, obtained from the time-resolved images, for a ROI drawn in the region corresponding to the myocardium; t_1 and t_2 are the beginning and end times of an acquisition frame (time bin); TBV and $(1 - TBV)$ (Hutchins GD *et al* J Nucl Med 1992; Muzic O *et al* J Nucl Med 1993) accounts for the SP contamination from

the LV blood pool to the myocardium at the early time points and PV recovery of the myocardium activity respectively, assuming that they add up to 1 and ρ is the myocardial tissue density (1.042 g of tissue/ml of tissue). $C_p(t)$ is the computed blood input curve and $C_T(t)$ is the model tissue, obtained by solving the differential equations indicating the rate of change of FDG concentration in the tissue as it traverses from the vascular to the extravascular and then to the cellular spaces (Zhong and Kundu 2013), defined as follows:

$$C_T(t) = \frac{K_1}{a_1 - a_2} \cdot [(k_3 + k_4 - a_1)e^{-a_1 t} + (a_2 - k_3 - k_4)e^{-a_2 t}] \otimes C_p(t) \quad (2),$$

where K_1 - k_4 are the kinetic parameters of the 3-compartment model (Zhong and Kundu 2013), \otimes is the convolution operator and

$$a_{2,1} = \left(\frac{1}{2}\right)(k_2 + k_3 + k_4 \pm \sqrt{(k_2 + k_3 + k_4)^2 - 4k_2k_4}) \quad (3)$$

Substituting equation (2) in (1) and assuming $k_4=0$, as FDG gets trapped as FDG-6-phosphate and does not de-phosphorylate, we obtain the following:

$$C_m(t) = \frac{1}{(t_2 - t_1)} \int_{t_1}^{t_2} \{(1 - TBV)\rho \left[\frac{K_1 k_3}{k_2 + k_3} \int_0^T C_p(u) du + \frac{K_1 k_2}{k_2 + k_3} \int_0^T C_p(u) \cdot e^{-(k_2 + k_3)(T-u)} du \right] + TBV \cdot C_p(T)\} dT \quad (4)$$

Equation (4) was numerically solved using non-linear regression analysis in MATLAB_r2018a (Mathworks Inc., Natick MA), to obtain K_1 - k_3 and hence K_i , defined as:

$$K_i = \frac{K_1 k_3}{(k_2 + k_3)} \quad (5)$$

Input curve from the LV cavity and model corrected blood input function

(LVBP-SP-PV): In addition to the IVC, ROIs were drawn in the regions corresponding to the LV cavity (median volume: 8.4 mm³) and myocardium and time activity curves generated. Using the formalism developed in mice (Zhong and Kundu 2013) and recently validated in rats (Li *et al*/JAHA 2019), wherein, the kinetic parameters and the model corrected blood input function (MCIF) with spill over and partial volume corrections are determined simultaneously by optimization of a 15-parameter objective function as described (Zhong and Kundu 2013), K_i was computed for WKY rat hearts.

2.5 Statistical Analysis:

All data were reported as mean \pm SD. Computed and measured K_i were compared using paired Student's *t* tests. A *p* value of less than 0.05 was considered statistically significant.

3. Results

In Figure 1a, example time-resolved FDG PET images of IVC in control WKY rats are shown. Here we see that FDG in blood washes in and out of IVC very rapidly. This resulted in underestimation of image derived blood input function (IDIF) especially at late time points. The arrows in the image point toward the IVC, heart and the brain. Regions of interest are drawn on IVC (blood) (Figure 1b) and adjacent or surrounding regions (tissue) (Figure 1c) and time activity curves generated (Figure 1d). Using a dual output model (Zhong and Kundu 2013) the spill-in contribution from the tissue to blood was computed to be only 5.2% and hence not included in our computation of K_i .

Partial volume correction was performed on IDIF obtained from IVC as shown in Figure 2 a–d. A RC plot was generated as a function of object size (1–7 mm) (Figure 2a) by performing a convolution of object size with a Gaussian distribution representing the spatial resolution of the Albira PET imaging system. Representative example 2D plot indicates original image of the simulated object (IVC) (left) (Seitz *et al* 2016) and that obtained after convolution (right) (Figure 2b). The 3D view of the intensity distributions are shown in Figure 2c, indicating a RC of ~70–80% for IVC after convolution considering detector blur. The average RC for eight control WKY rats were computed to be $75 \pm 13\%$. In Figure 2d, example uncorrected IDIF and PV corrected IDIF for a control WKY rat are shown.

In Figure 3a–b, example transverse images of IVC and LV are shown with regions of interest (ROI) drawn and time activity curves (TAC) generated for the ROI's. PV corrected IDIF from IVC (blood TAC) and model fit for tissue TAC correcting for the SP contamination from the LV blood pool to the myocardium at the early time points and the PV recovery of the myocardium in a 5-parameter 3-compartment kinetic model are shown in Figure 3c. The average K_i for control WKY rat hearts using input curve from IVC was computed to be 0.0409 ± 0.019 (ml/min/g). Ratio of the standard deviation to the mean indicated a variation of 46.4% in the computed K_i .

Representative time-resolved FDG PET images of the heart (Figure 4a) and regions of interest in the LV blood pool and the myocardium are shown in Figure 4b–c. Time activity curves obtained from the time-resolved images, computed model fits and model corrected blood input function (MCIF) are shown in Figure 4d. The average K_i for control KKY rat hearts using input curve from LV blood pool was computed to be 0.0479 ± 0.026 ml/min/g. Ratio of the standard deviation to the mean indicated a variation of 55.7% in the computed K_i .

There is no significant difference between the computed K_i obtained from IVC and LVBP ($p=0.56$). The computed K_i using input curve from IVC (IVC-PV) and LVBP (LVBP-SP-PV) were then compared to that measured from arterial blood samples in control SD rats (0.042 ± 0.022) (Fang and Muzic 2008). There was no significant difference observed between computed and measured K_i , using IVC-PV ($p=0.91$) and LVBP-SP-PV ($p=0.76$).

4. Discussion

The standard of care for obtaining blood input function in dynamic FDG PET imaging of rodents is arterial blood sampling (Laforest *et al* 2005, Pain *et al* 2004). This method is invasive (by catheterization of the carotid or femoral arteries) and requires significant blood sampling, more so at the early time points at a rapid rate (~10s intervals for a minute to catch the peak and coarser time intervals after that for 60 minutes to match the time bins obtained from imaging) due to quick changes in FDG metabolism at early time. Apart from the challenging nature of the surgical procedure for arterial blood sampling, due to limited blood volume of rodent hearts, significant blood sampling may compromise the physiological condition of the heart, there by affecting the resultant measured myocardial FDG uptake rates.

In order to minimize blood sampling, several methods have been developed over the years by us and others, which relies on image derived blood input function (IDIF), for computation of myocardial FDG uptake rates, K_i , *in vivo*. Hybrid methods (Shoghi and Welch 2007) rely on IDIF at the early time points and blood samples at the late time points to compute K_i . Due to the limited intrinsic spatial resolution of the rodent scanners, IDIF and tissue time activity curves are susceptible to PV averaging and SP contamination both at early and late time points in a dynamic PET scan. Prior methods by Fang et al (Fang and Muzic 2008) developed a parametric formulation of the output equation and optimized with the TACs derived from the FBP reconstructed dynamic PET images, which resulted in simultaneous estimation of model corrected input function (MCIF) and the kinetic parameters *in vivo*. Work performed in our lab concluded that high resolution reconstruction algorithm and cardiac gating can minimize SP contamination from the myocardium to the LV blood pool at the late time points, resulting in K_i not significantly different from that obtained using arterial blood sampling (Locke *et al* 2011). However, our results indicated that the above method did not completely eliminate SP contamination (~15%) at the late time points confounding its effect for smaller heart sizes. Therefore, a recent work from our lab combining high resolution reconstruction methods with parametric optimization and cardiac gating eliminated SP contamination (~5%) resulting in significant improvement in computed K_i when compared to that obtained from arterial blood sampling (Zhong and Kundu 2013). The optimization method here used “fmincon” in Matlab, which utilizes a deterministic algorithm based on Interior Reflective Newton (IRN) algorithm to search for a local minimum. Another recent work from our lab combined stochastic and deterministic methods to search for global minimum, resulting in improve estimation of the computed K_i in a 15-parameter 3-compartment model (Li and Kundu 2018). However, due to the multi-parameter nature of these methods, the computed K_i is sensitive to bounds of the model parameters and to initial guess values. Further work by others (Kim *et al* 2006, Fakhri *et al* 2005) and us (Huang *et al* 2018) used factor analysis techniques using principle component analysis (PCA) to segment out LV blood pool from the myocardium and the surrounding background to compute IDIF based on time-dependent pixels in dynamic FDG PET scans of WKY rat hearts. This method, although requires no human intervention in defining ROI's, underestimates computed K_i by a factor of ~2 when compared to that obtained from arterial

blood samples, due to significant SP-in contamination from the myocardium to the LV blood pool at the late time points.

Recent works utilized IVC as a source of the blood input function for computing Ki, as FDG washes in and out of IVC rapidly. Also since there is no tissue around IVC, SP contamination into blood is completely eliminated. However, due to the limited size of IVC, the IDIF is susceptible to incomplete radioactivity recovery or PV averaging. This results in underestimation of IDIF at the late time points and thereby overestimation of the computed Ki. One of the works (Lanz *et al* 2014) corrected for dispersion due to difference in sampling sites of the blood input curve (IVC opposed to LV blood pool) but had to rely on blood sampling for this correction to compute cerebral FDG uptake rates. Another work (Thackeray *et al* 2015) in mouse hearts did not perform PV or dispersion corrections and concluded IVC as a reliable source of obtaining Ki compared to LV blood pool in a graphical Patlak model (Patlak and Blasberg 1985).

Our work is different from the above in that we performed PV corrections to IDIF obtained from IVC non-invasively without correcting for dispersion effects in a 3-compartment kinetic model with SP corrections from the blood to the myocardium at the early time points and also correcting for PV averaging of the myocardium, for the first time, in WKY rat hearts *in vivo*. This resulted in ~46% deviation of the computed Ki from the mean and not significantly different from that measured using arterial blood sampling in control SD rat hearts. These computations however do not include SP contamination from adjacent regions (tissue) to the blood (IVC) at the late time points, which is negligible as shown in Figures 1b-d.

We used the metabolic tracer FDG to first validate our computation model in control WKY rat, which was then applied to sensitively detect failing versus non-failing experimental spontaneously hypertensive rat (SHR) hearts *in vivo* (Huang *et al* 2018). However, it should be noted that the computation model may be easily adapted to other tracers as the kinetics of FDG is very similar to ¹³N-ammonia or ⁸²Rb.

Computations using input curve from the LVBP (LVBP-SP-PV), considered as the standard method (Zhong and Kundu 2013), in the same WKY rats in a 15-parameter model with SP and PV corrections resulted in a mean Ki not different from that using IVC with partial volume corrections (IVC-PV) and also not significantly different from that obtained using arterial blood sampling in SD rats (Fang and Muzic 2008). The multi-parameter nature of the optimization problem using LVBP-SP-PV, however, resulted in a higher deviation in the mean computed Ki compared to that obtained using the IVC (IVC-PV).

The method of deriving the input curve from the IVC and correcting for PV in a 5-parameter 3-compartment model maybe applicable for all dynamic FDG PET studies to compute Ki and maybe more robust compared to that obtained from LVBP in rats and mice. However, in case of applications with tracers wherein the IVC cannot be clearly discerned from the PET images and the input curve needs to be derived from the LVBP, the 15-parameter model is more appropriate especially in studies where we have smaller structure of the heart and SP contamination from the myocardium to the LVBP becomes significant.

Our study is not without limitations. We compare our downstream Ki to that obtained using arterial blood sampling in control SD rats. This is a limitation of the study as there is no study which has performed arterial sampling in control WKY rats. However, we expect the glucose metabolic rates to be similar in these strains of control rats. Studies in Zucker lean rats [control of Zucker Diabetic Fatty (ZDF) rats], which computed rates of myocardial glucose uptake and utilization, indicated a similar Ki range (Shogi *et al* 2008). Another study in healthy Fischer rats weighing around 200 g (Mabrouk *et al* 2013), based on their computed rates of myocardial glucose utilization, also indicated a similar Ki range. In a recent review article (Cicone *et al* 2017), the authors identified control WKY and SD rats as commonly used strains in both cardiac PET and SPECT imaging studies. In the future, we will perform dynamic FDG PET imaging and arterial blood sampling at the same time in control WKY rats for direct (IDIF) and indirect (Ki) comparison of the computed methods (from IVC and LVBP).

5. Conclusions

In conclusion, we find that the input curve obtained from the IVC corrected for PV averaging is comparatively more repetitive to that obtained from the LV cavity in that Ki obtained with the former exhibits 46% mean deviation in the computed values compared to 56% with the latter in control WKY rat hearts. The computed Ki using the input curve from IVC with PV corrections and LVBP with SP and PV corrections, however, compare well with each other and with measured Ki using arterial blood samples in control SD rats.

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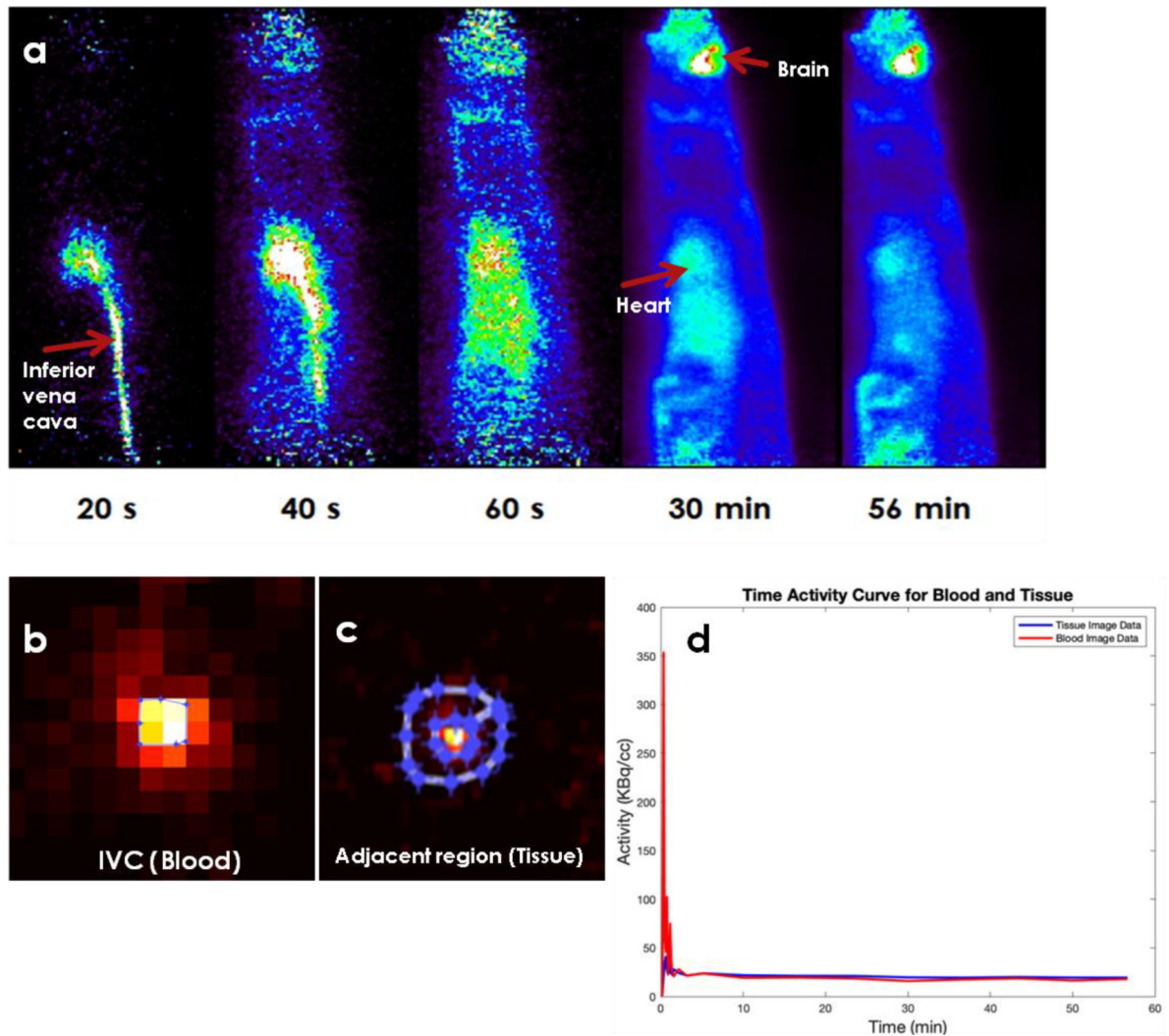


Figure 1: Example time-resolved FDG PET images of IVC in control WKY rats.

a) Time-resolved FDG PET images of the inferior vena cava (IVC) over a period of 60 minutes. The images indicate rapid wash-in and wash-out of FDG from IVC. The arrows point to the inferior vena cava, heart and brain. **b)** Regions of interest drawn on IVC (zoomed in), **c)** adjacent region and **d)** time activity curves for IVC and adjacent region. Computed adjacent region (tissue, t) to IVC (blood, b) spill-in (S_{tb}) contamination using formalism developed earlier (Zhong and Kundu 2013) was negligible ($\sim 5\%$) and hence not included in the computation of K_i .

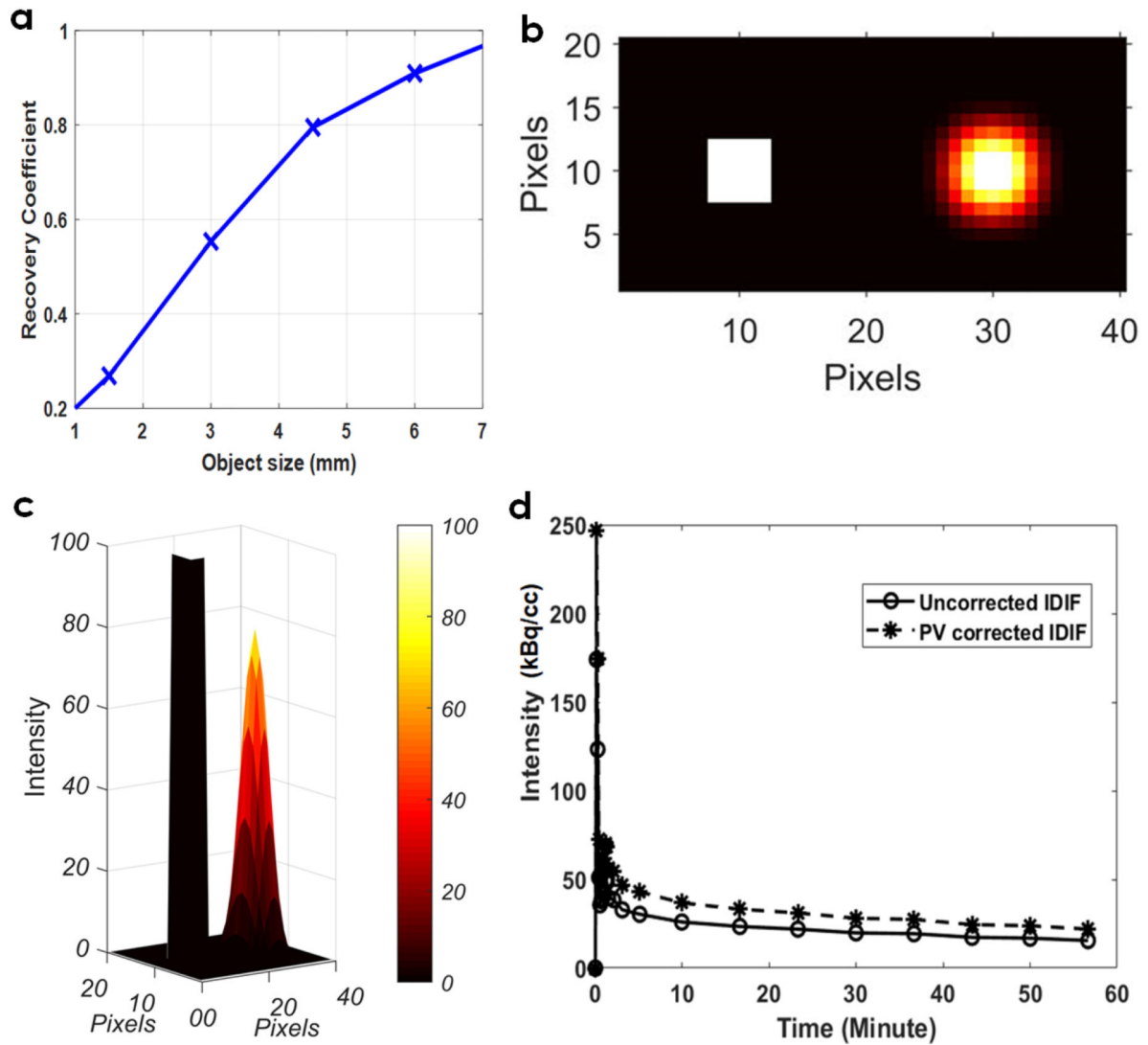


Figure 2: Partial volume corrections on IDIF obtained from IVC.

(a) Recovery coefficient (RC) plot as a function of object size (1–7 mm) obtained by convolution of object size with a Gaussian distribution representing the spatial resolution of the Albira PET system. (b) Representative example 2D plot indicates image of the simulated object (IVC) (left) and that obtained after convolution (right). (c) 3D view of the intensities distributions are shown, indicating a RC of ~70–80% for IVC. (d) Example uncorrected IDIF and PV corrected IDIF for a control WKY rat.

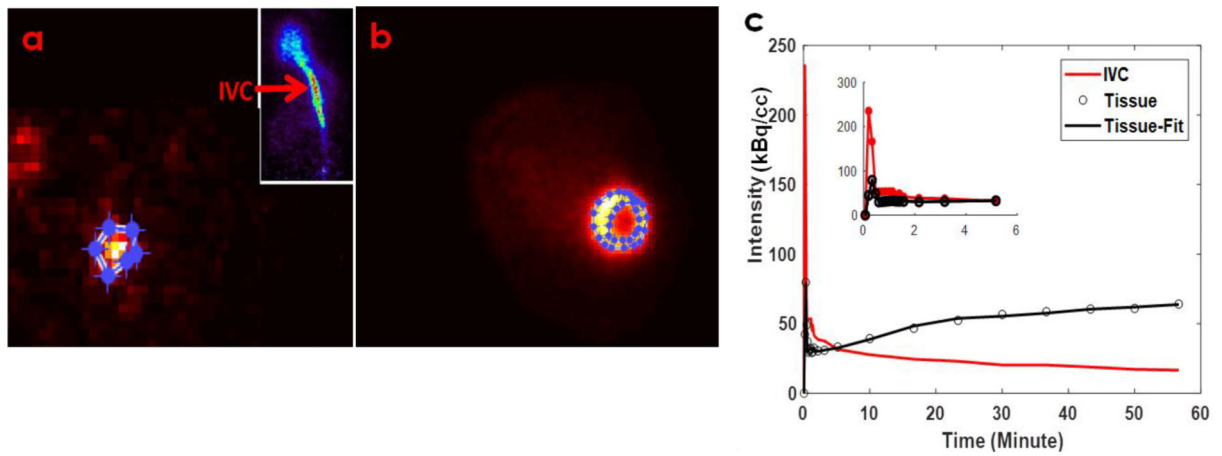


Figure 3: FDG PET images of the IVC, LV and model fit.

Example transverse images of IVC (**a**) and LV (**b**) are shown with regions of interest (ROI) drawn and time activity curves (TAC) generated for the ROI's. (**c**) PV corrected IDIF from IVC (blood TAC; solid red line) and model fit (solid black line) for tissue TAC (open circles) correcting for SP contamination from the LV blood pool to the myocardium at the early time points and PV recovery of the myocardium in a 5-parameter 3-compartment kinetic model.

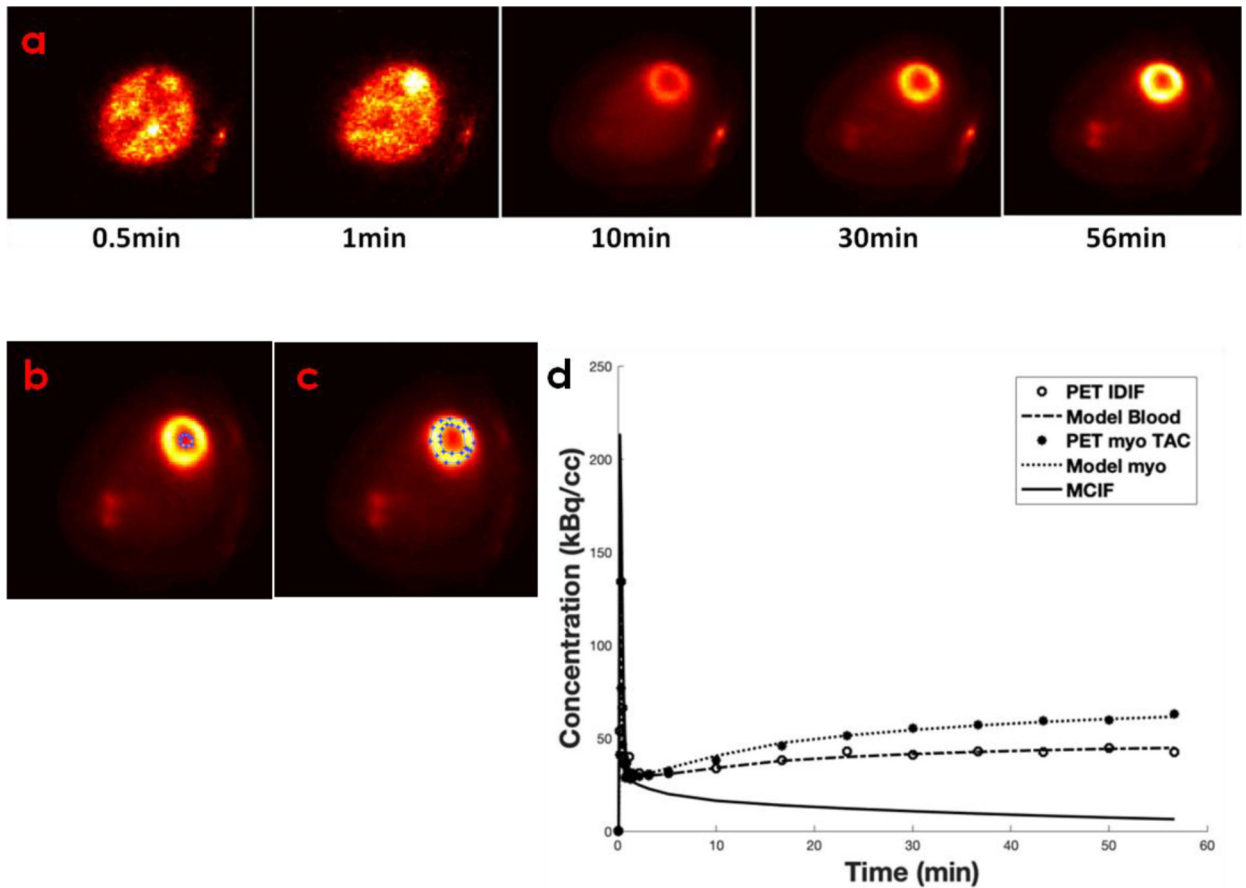


Figure 4: Time resolved FDG PET images of the heart and model corrected blood input function (MCIF).

(a) Example dynamic FDG PET images of WKY rat heart *in vivo* over a period of 60 minutes. (b) Regions of interest drawn in the LV blood pool and (c) the myocardium. (d) Representative time activity curves for LV blood pools and myocardium, model fits, and model corrected blood input function (MCIF).

Table 1

Abbreviations

PET	Positron Emission Tomography
FDG	2-[18F] fluoro-2-deoxy-D-glucose
FOV	Field of View
LM	List-mode
MLEM	Maximum Likelihood Expectation Maximization
IVC	Inferior Vena Cava
LVBP	Left Ventricular Blood Pool
PV	Partial Volume
SP	Spill-over
IVC-PV	IVC corrected for PV
LVBP-SP-PV	LVBP corrected for SP and PV
IDIF	Image Derived Input Function
MCIF	Model Corrected Input Function
ROI	Regions of Interest
Ki	Rate of myocardial FDG Uptake
WKY	Wistar-Kyoto
SD	Sprague Dawley

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