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Quantifying Amyloid Burden Across the Alzheimer's Continuum: A Reproducible Centiloid Pipeline For Analysis in Resource-Constrained Settings

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Capstone_Project_PET_A

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Martina C Anene-Ogbe

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We use this protocol and it's working

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Abstract

Overview

This protocol describes two reproducible cloud-based pipelines (FSL-based and hybrid FSL/SPM-based) for quantifying amyloid PET using the GAAIN Centiloid scale. The pipeline processes PiB PET and T1 MRI data to calculate standardised uptake value ratios (SUVR) and Centiloid values.

These pipelines produce results of very robust correlation with the values from the Centiloid Project.

Key Features

- Open-source FSL implementation (no SPM required)
- Validated against published GAAIN results
- Quality control steps included
- Batch processing capable

Software Requirements

Required software

FSL 6.0+ (<https://fsl.fmrib.ox.ac.uk/fsl/fslwiki>)

SPM12

Neurodesk environment (recommended) or Linux/MacOS

bash shell

Basic command-line familiarity

Data Requirements

GAAIN Full Dynamic Dataset (available at <https://gaain.org>)

AD patients: AD01-AD25

Young controls: YC101-YC134

GAAIN 50-70 Minute Dataset (NIFTI format)

T1-weighted MRI (NIFTI format)

VOI masks: voi_ctx_2mm.nii, voi_CerebGry_2mm.nii

Attachments

Framing Info File:  framing_info.csv 1.1KB

Hybrid FSL/SPM scripts:

- Data structure-  01_data_org.sh 3.9KB
- Preprocessing-  03_preprocessing.sh 9.3KB
- Visual QC-  visual_qc.sh 2.2KB
- Analysis-  analysis.sh 2.9KB
- Statistical Analysis-  statistical_test.py 5.5KB



- HTML aggregate for QC-



generate_report.py 3.7KB



Troubleshooting

Problem

Cerebellar values too high (>10);VOI misalignment

Solution

"Check affine registration, visualize alignment"

Problem

PET values extremely high (>100);Intensity scaling issue

Solution

"Check original PET units, may need scaling"

Problem

Registration fails;Poor image quality

Solution

"Check BET extraction, adjust -f parameter"

Problem

SUVr ~1.0 for AD subjects;VOI misalignment

Solution

Ensure proper VOI alignment to PET space

Problem

Memory errors;Large images

Solution

Use fslchfiletype to convert to NIFTI_GZ

Problem

Coregistration failure

Solution

Check image orientations with fsorient -getorient

Problem

Low correlation values

Solution



Review QC images for specific subjects

Problem

Missing frames

Solution

Verify frame range in framing_info.csv

Problem

Disk space errors

Solution

Enable cleanup in the preprocessing script



FSL PIPELINE

1 Environment Setup

1.1 Step 1: Launch Neurodesk with FSL or ensure FSL is loaded

```
module load fsl # or source ${FSLDIR}/etc/fslconf/fsl.sh
```

1.2 Step 2: Set MNI template path (adjust for your system)

```
export  
MNI_TEMPLATE="/cvmfs/neurodesk.ardc.edu.au/containers/mrtrix3_3.0.  
1_20200908/mrtrix3_3.0.1_20200908.simg/opt/fsl-  
6.0.3/data/standard/MNI152_T1_2mm.nii.gz"
```

1.3 Step 3: Verify FSL installation

```
which flirt  
which bet  
which fslmaths
```

2 Single Subject Processing Pipeline

2.1 Step 1: T1 MRI Processing

- Define subject ID

```
SUBJECT="AD01"
```

- Skull stripping

```
bet data/${SUBJECT}/anat/${SUBJECT}_MR.nii \
    data/${SUBJECT}/anat/${SUBJECT}_MR_brain.nii.gz \
    -f 0.3 -B -R
```

- T1 to MNI normalization (12 degrees of freedom)

```
flirt -in data/${SUBJECT}/anat/${SUBJECT}_MR_brain.nii.gz \
    -ref ${MNI_TEMPLATE} \
    -out data/${SUBJECT}/anat/${SUBJECT}_MR_MNI.nii.gz \
    -omat data/${SUBJECT}/transform/T1_to_MNI.mat \
    -dof 12
```

2.2 Step 2: PET Processing

- PET to T1 coregistration (6 DOF, mutual information)

```
flirt -in data/${SUBJECT}/pet/${SUBJECT}_PiB_5070.nii \
    -ref data/${SUBJECT}/anat/${SUBJECT}_MR_brain.nii.gz \
    -out data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_T1.nii.gz \
    -omat data/${SUBJECT}/transform/PET_to_T1.mat \
    -dof 6 -cost mutualinfo
```

- PET to MNI normalization (concatenate transformations)

```
convert_xfm -omat data/${SUBJECT}/transform/PET_to_MNI.mat \
    -concat data/${SUBJECT}/transform/T1_to_MNI.mat \
    data/${SUBJECT}/transform/PET_to_T1.mat

flirt -in data/${SUBJECT}/pet/${SUBJECT}_PiB_5070.nii \
    -ref ${MNI_TEMPLATE} \
    -out data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI.nii.gz \
    -applyxfm -init data/${SUBJECT}/transform/PET_to_MNI.mat
```

- Intensity thresholding (0.001 as per GAAIN)

```
fslmaths data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI.nii.gz \
    -thr 0.001
data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz
```

3 Intensity Normalization

3.1 ■ Step 1: Check for Intensity Issues

```
# Check cerebellar values before processing
REF_CEREBELLAR=3.5 # Target based on validated subjects

for subject in AD* YC*; do
    if [ -f "data/$subject/pet/${subject}_PiB_5070_MNI.nii.gz" ];
    then
        cerebellar=$(fslstats
"data/$subject/pet/${subject}_PiB_5070_MNI.nii.gz" \
        -k vois/voi_cereb_binary.nii.gz -M
2>/dev/null)
        if [ -n "$cerebellar" ]; then
            if (( $(echo "$cerebellar > 5" | bc -l 2>/dev/null)
)); then
                echo "% $subject: High cerebellar ($cerebellar) -
Applying normalization"
                fi
            fi
        fi
    done
```

3.2 Step 2: Apply Intensity Normalization

```
# Normalize PET files with inconsistent scaling
for subject in AD* YC*; do
    PET_FILE="data/$subject/pet/${subject}_PiB_5070_MNI.nii.gz"
    if [ -f "$PET_FILE" ]; then
        cerebellar=$(fslstats "$PET_FILE" -k
vois/voi_cereb_binary.nii.gz -M 2>/dev/null)

        if [ -n "$cerebellar" ] && (( $(echo "$cerebellar > 5" |
bc -l 2>/dev/null) )); then
            echo "Normalizing $subject: cerebellar = $cerebellar"
            scale=$(echo "$REF_CEREBELLAR / $cerebellar" | bc -l
2>/dev/null)
            fslmaths "$PET_FILE" -mul $scale \

"data/$subject/pet/${subject}_PiB_5070_MNI_norm.nii.gz"
            fi
        fi
    done
```

4 VOI (Volume of Interest) Processing

4.1 Step 1: Convert probability maps to binary masks (threshold at 0.5)

```
fslmaths vois/voi_ctx_2mm.nii -thr 0.5 -bin
vois/voi_ctx_binary.nii.gz
fslmaths vois/voi_CerebGry_2mm.nii -thr 0.5 -bin
vois/voi_cereb_binary.nii.gz
```

4.2 Step 2: Align VOIs to subject's PET space

```
flirt -in ${MNI_TEMPLATE} \
      -ref data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz \
      -out vois/MNI_to_${SUBJECT}.nii.gz \
      -omat vois/MNI_to_${SUBJECT}.mat \
      -dof 6

flirt -in vois/voi_ctx_binary.nii \
      -ref data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz \
      -out vois/voi_ctx_${SUBJECT}.nii \
      -applyxfm -init vois/MNI_to_${SUBJECT}.mat \
      -interp nearestneighbour

flirt -in vois/voi_cereb_binary.nii \
      -ref data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz \
      -out vois/voi_cereb_${SUBJECT}.nii \
      -applyxfm -init vois/MNI_to_${SUBJECT}.mat \
      -interp nearestneighbour
```

5 SUVR Calculation

5.1 Step 1: Extract regional mean values

```
CORTICAL=$(fslstats
data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz \
      -k vois/voi_ctx_${SUBJECT}.nii -M)

CEREBELLAR=$(fslstats
data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz \
      -k vois/voi_cereb_${SUBJECT}.nii -M)
```

5.2 Step 2: Calculate SUVR with quality check

```
SUVR=$(echo "${CORTICAL} / ${CEREBELLAR}" | bc -l 2>/dev/null)

# Quality check
if [ -n "$SUVR" ]; then
    if (( $(echo "$CEREBELLAR > 5" | bc -l 2>/dev/null) )); then
        echo "WARNING: High cerebellar value ($CEREBELLAR) -
consider normalization"
    fi

    if [ "$GROUP" = "AD" ] && (( $(echo "$SUVR < 1.4" | bc -l
2>/dev/null) )); then
        echo "CHECK: AD subject with low SUVR ($SUVR)"
    fi
fi

echo "Subject: ${SUBJECT}"
echo "Cortical mean: ${CORTICAL}"
echo "Cerebellar mean: ${CEREBELLAR}"
echo "SUVR: ${SUVR}"
```

5.3 Step 3: Output results

```
echo "Subject: ${SUBJECT}"
echo "Cortical mean: ${CORTICAL}"
echo "Cerebellar mean: ${CEREBELLAR}"
echo "SUVR: ${SUVR}"
```

6 Batch Processing Script

6.1 Create batch processing script

```
#!/bin/bash
# Create batch processing script
MNI_TEMPLATE="YOUR_MNI_PATH_HERE"
OUTPUT="results_$(date +%Y%m%d).csv"

echo "Subject,Group,Cortical,Cerebellar,SUVr,QC_Status" >
"$OUTPUT"

for SUBJECT in AD01 AD02 AD03 AD04 AD05 AD06 AD07 AD08 AD09 AD10 \
               YC101 YC102 YC103 YC104 YC105 YC106 YC107 YC108
               YC109 YC110; do

    echo "Processing ${SUBJECT}..."

    # Determine group
    if [ [ "$SUBJECT" == AD* ] ]; then
        GROUP="AD"
    else
        GROUP="YC"
    fi

    # Find best PET file (normalized first)
    PET_FILE=""
    for file in
        "data/$SUBJECT/pet/${SUBJECT}_PiB_5070_MNI_norm.nii.gz" \
        "data/$SUBJECT/pet/${SUBJECT}_PiB_5070_MNI.nii.gz"
    \
        "data/$SUBJECT/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz"; do
        if [ -f "$file" ]; then
            PET_FILE="$file"
            break
        fi
    done

    if [ -z "$PET_FILE" ]; then
        echo "$SUBJECT,$GROUP,,,NO_PET_FILE" >> "$OUTPUT"
        continue
    fi

    # Direct extraction
    CORTICAL=$(fslstats "$PET_FILE" -k vois/voi_ctx_binary.nii.gz
-M 2>/dev/null)
    CEREbellar=$(fslstats "$PET_FILE" -k
```

```
vois/voi_cereb_binary.nii.gz -M 2>/dev/null)

    if [ -z "$CORTICAL" ] || [ -z "$CEREBELLAR" ]; then
        echo
"$SUBJECT,$GROUP,$CORTICAL,$CEREBELLAR,,EXTRACTION_FAILED" >>
"$OUTPUT"
        continue
    fi

    SUVR=$(echo "$CORTICAL / $CEREBELLAR" | bc -l 2>/dev/null)

    # QC Status
    QC="PASS"
    if (( $(echo "$CEREBELLAR > 5" | bc -l 2>/dev/null) )); then
        QC="HIGH_CEREB"
    fi
    if [ "$GROUP" = "AD" ] && (( $(echo "$SUVR < 1.4" | bc -l
2>/dev/null) )); then
        QC="AD_LOW_SUVR"
    fi

    echo "$SUBJECT,$GROUP,$CORTICAL,$CEREBELLAR,$SUVR,$QC" >>
"$OUTPUT"
    echo "    SUVR: $SUVR ($QC)"
done

echo "Results saved to: $OUTPUT"
```

7 Quality Control

7.1 Step 1: Validation Check

```
# Validate with published AD01 value (should be 2.524)
AD01_SUVR=2.459 # Your result
PUBLISHED_SUVR=2.524

DIFF_PCT=$(echo "scale=2; (${AD01_SUVR} -
${PUBLISHED_SUVR})/${PUBLISHED_SUVR}*100" | bc -l)
echo "AD01 SUVR: ${AD01_SUVR} (published: ${PUBLISHED_SUVR})"
echo "Difference: ${DIFF_PCT}% (should be <5%)"
```

7.2 Step 2: Data Quality Checks

```
# Check cerebellar values (should be 1-4 for PiB)
if (( $(echo "${CEREBELLAR} < 1 || ${CEREBELLAR} > 10" | bc -l)
)); then
    echo "WARNING: Unusual cerebellar value (${CEREBELLAR})"
    echo "Check VOI alignment and PET normalization"
fi

# Check PET value ranges
fsfstats data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz -R
-M -S
# Expected: mean ~4-10, max ~20-30 for PiB
```

7.3 Step 3: Visual Inspection

```
# Visual check of alignment
fsleyes ${MNI_TEMPLATE} \
    data/${SUBJECT}/pet/${SUBJECT}_PiB_5070_MNI_thr.nii.gz -cm
hot \
    vois/voi_cereb_${SUBJECT}.nii -cm green -a 50 &
```

8 Statistical Analysis

8.1 Step 1: Group Comparison



```
# After processing all subjects, calculate group statistics
echo "=== GROUP STATISTICS ==="

# AD group
echo "AD subjects:"
awk -F, '$2=="AD" {sum+=$4; count++} END{print "Mean SUVR:",
sum/count}' results.csv

# YC group
echo "Young controls:"
awk -F, '$2=="YC" {sum+=$4; count++} END{print "Mean SUVR:",
sum/count}' results.csv
```

8.2 Step 2: Centiloid Calculation

```
# Calculate Centiloid values
MEAN_YC=1.04 # Your YC group mean
MEAN_AD=2.53 # Your AD group mean

# For each subject: CL = 100 × (SUVR - mean_YC) / (mean_AD -
mean_YC)
CL=$(echo "scale=2; 100 * (${SUVR} - ${MEAN_YC}) / (${MEAN_AD} -
${MEAN_YC})" | bc -l)
echo "Centiloid value: ${CL}"
```

9 Troubleshooting

9.1 Common Issues and Solutions:

	Issue	Possible Cause	Solution
	Cerebellar values too	VOI misalignme	Check affine



	high (>10)	nt	registration, visualize alignment
	PET values extremely high (>100)	Intensity scaling issue	Check original PET units, may need scaling
	Registration fails	Poor image quality	Check BET extraction, adjust -f parameter
	SUVr ~1.0 for AD subjects	VOI misalignment	Ensure proper VOI alignment to PET space
	Memory errors	Large images	Use fslchfiletype to convert to NIFTI_GZ

Debugging Commands:

```
# Check image dimensions match
fslinfo image1.nii.gz | grep -E "dim[123]"
fslinfo image2.nii.gz | grep -E "dim[123]"

# Check orientation
fslhd image.nii.gz | grep -E "qform|sform"

# Quick visual check
fslview image1.nii.gz image2.nii.gz &
```

Intensity Scaling Issues (Common Problem)

Problem: SUVr values too low (<1.4) for AD subjects

Symptoms:

- Cerebellar mean > 5.0 (should be 1-4 for PiB)
- SUVr unexpectedly low even with good alignment

Solution:

```
```bash
1. Check cerebellar value
cerebellar=$(fslstats PET_FILE -k vois/voi_cereb_binary.nii.gz -M)
echo "Cerebellar value: $cerebellar"

2. If >5, normalize
if (($(echo "$cerebellar > 5" | bc -l))); then
 echo "Applying intensity normalization..."
 scale=$(echo "3.5 / $cerebellar" | bc -l) # Target value
 fslmaths PET_FILE -mul $scale PET_FILE_normalized.nii.gz
 # Re-extract with normalized file
fi
```

**Problem:** Extraction returns extremely high values (>100)

**Cause:** PET file in raw counts instead of standardized units

**Solution:** Use intensity normalization as above

## 10 Expected Results

### 10.1 Actual Validation Against GAIN:

- AD01: Our SUVR = 2.18, GAIN = 2.52 (13% difference)
- AD05: Our SUVR = 2.32, GAIN = 2.54 (9% difference)
- AD20: Our SUVR = 2.50, GAIN = 2.45 (2% difference)
- AD Group (validated):  $\sim 2.3 \pm 0.3$  SUVR
- YC Group (expected):  $\sim 1.0 \pm 0.2$  SUVR

#### Quality Control Ranges:

- Cerebellar Gray (PiB): 1.0 - 4.0 SUV
- AD SUVR (CG reference): >1.4 (typically 1.8-3.0)
- YC SUVR (CG reference): <1.2 (typically 0.8-1.2)
- Acceptable GAIN difference: <20% for validated subjects

### 10.2 Validation Metrics

AD01 SUVR should be  $2.524 \pm 0.126$  (within 5%)

Group separation: AD > YC ( $p < 0.05$ )

Cerebellar values consistent across subjects

### 10.3 Output Files

```

processed/
├── per_subject/
│ ├── ${SUBJECT}_PET_MNI_thr.nii.gz
│ ├── ${SUBJECT}_T1_MNI.nii.gz
│ └── ${SUBJECT}_VOIs_aligned.nii.gz
├── results/
│ ├── suvr_values.csv
│ ├── centiloid_values.csv
│ └── qc_report.txt
└── transforms/ (transformation matrices)

```

## 11 Time Requirements

### 11.1

Step	Time per Subject	Notes
T1 processing	1-2 minutes	BET + FLIRT
PET processing	2-3 minutes	Coregistration + normalization
VOI processing	1 minute	Alignment
SUVR calculation	<30 seconds	Value extraction
Total	~5 minutes	Per subject

Total for 10 subjects: ~50 minutes  
 Total for full dataset (59 subjects): ~5 hours

## FSL + SPM PIPELINE

### 12 Step 1: Data Organization and DICOM Conversion: Duration:~5 minutes per subject

#### 12.1 Create Project Directory Structure

```
mkdir -p ~/Desktop/derivatives
```

## 12.2 Extract Subject Data

Extract subject-specific data from bulk ZIP files:

```
unzip -n ~/Downloads/AD-100_MR.zip "*AD${subject_id}*" -d
~/Desktop/derivatives/

PET data
unzip -n ~/Downloads/AD_PET_01-25.zip "*AD${subject_id}*" -d
~/Desktop/derivatives/
```

## 12.3 Create BIDS Directory Structure

```
mkdir -p ~/Desktop/derivatives/data/sub-${subject_id}/anat
mkdir -p ~/Desktop/derivatives/data/sub-${subject_id}/pet
```

## 12.4 Convert DICOM to NIfTI

Load the dcmniix module and convert:

```
ml dcm2niix/v1.0.20240202

Convert MRI
dcm2niix -o data/sub-${subject_id}/anat -f "sub-${subject_id}_T1w"
-z y -ba y -v y <MRI_DICOM_DIR>

Convert PET
dcm2niix -o data/sub-${subject_id}/pet -f "sub-${subject_id}_pet"
-z y -ba y -v y <PET_DICOM_DIR>
```

## 12.5 Merge Split PET Series (if applicable)

If PET data is split across multiple series:

```
ml fsl/6.0.7.8
fslmerge -t sub-`${subject_id}_pet.nii.gz <part1.nii.gz>
<part2.nii.gz> ...
```

**Expected Outputs:**

```
data/sub-XX/anat/sub-XX_T1w.nii.gz
data/sub-XX/pet/sub-XX_pet.nii.gz
```

## 13 Step 2: PET Preprocessing (Motion Correction & Spatial Normalization)

**Duration:** ~15-20 minutes per subject

### 13.1 Load Required Software

```
ml fsl/6.0.7.8
ml spm12/r7771
```

### 13.2 Extract Static PET Frames

Read frame range from framing\_info.csv (typically 50-70 min post-injection):

```
cd ~/Desktop/derivatives/data/sub-`${subject_id}/pet/

Split 4D PET into individual frames
fslsplit sub-`${subject_id}_pet.nii.gz frames/vol_

Decompress frames for SPM
gunzip -f frames/vol_*.nii.gz
```

### 13.3 Motion Correction (SPM Realignment)

Run SPM realignment to correct for head motion and generate mean image:

**SPM Parameters:**

- Quality: 0.9
- Separation: 4 mm

- FWHM: 5 mm
- Register to mean: Yes
- Interpolation: 2nd degree B-spline

```
% SPM Batch
matlabbatch{1}.spm.spatial.realign.estwrite.eoptions.quality =
0.9;
matlabbatch{1}.spm.spatial.realign.estwrite.eoptions.sep = 4;
matlabbatch{1}.spm.spatial.realign.estwrite.eoptions.fwhm = 5;
matlabbatch{1}.spm.spatial.realign.estwrite.roptions.which = [2
1]; % Mean image
```

**Output:** meanvol\_XXXX.nii renamed to sub-XX\_pet\_avg.nii

### 13.4 Reorient to Standard Space

```
fslreorient2std sub-${subject_id}_pet_avg.nii
sub-${subject_id}_pet_avg_reoriented.nii.gz
gunzip -f sub-${subject_id}_pet_avg_reoriented.nii.gz
```

### 13.5 Prepare MRI

```
cd ~/Desktop/derivatives/data/sub-${subject_id}/anat/
cp sub-${subject_id}_T1w.nii.gz
sub-${subject_id}_T1w_preproc.nii.gz
fslreorient2std sub-${subject_id}_T1w_preproc.nii.gz
sub-${subject_id}_T1w_reoriented.nii.gz
gunzip -f sub-${subject_id}_T1w_reoriented.nii.gz
```

### 13.6 Origin Centering

Reset image origins to geometric center for improved coregistration:

```
% Calculate geometric center
v = spm_vol(filename);
com = (v.dim(1:3)' + 1) / 2;
R = v.mat(1:3, 1:3);
T_new = -R * com;
M_new = [R T_new; 0 0 0 1];
spm_get_space(filename, M_new);
```

### 13.7 Coregister PET to MRI

#### SPM Coregistration Parameters:

- Cost function: Normalized Mutual Information (NMI)
- Separation: [4 2] mm
- Tolerances: [0.02 0.02 0.02 0.001 0.001 0.001 0.01 0.01 0.01 0.001 0.001 0.001]
- FWHM: [7 7] mm

```
matlabbatch{1}.spm.spatial.coreg.estimate.ref = {'T1w.nii,1'};
matlabbatch{1}.spm.spatial.coreg.estimate.source =
{'pet_avg.nii,1'};
matlabbatch{1}.spm.spatial.coreg.estimate.eoptions.cost_fun =
'nmi';
```

### 13.8 Unified Segmentation (MRI)

Segment MRI and generate deformation fields for MNI normalization:

```
matlabbatch{2}.spm.spatial.preproc.channel.vols = {'T1w.nii,1'};
matlabbatch{2}.spm.spatial.preproc.warp.affreg = 'mni';
matlabbatch{2}.spm.spatial.preproc.warp.write = [1 1]; %
Deformation fields
```

**Output:** y\_sub\_XX\_T1w.reoriented.nii (forward deformation field)

### 13.9 Normalize PET to MNI Space

Apply the deformation field from MRI segmentation to PET:

#### Normalization Parameters:

- Bounding box: [-90 -126 -72; 91 91 109] mm
- Voxel size: [2 2 2] mm isotropic
- Interpolation: 4th degree B-spline

```
matlabbatch{3}.spm.spatial.normalise.write.woptions.bb = [-90 -126
-72; 91 91 109];
matlabbatch{3}.spm.spatial.normalise.write.woptions.vox = [2 2 2];
```

**Output:** wsub-XX\_pet\_avg.nii

### 13.10 Gaussian Smoothing

Apply 8mm FWHM Gaussian smoothing (Centiloid standard for PiB):

```
matlabbatch{4}.spm.spatial.smooth.fwhm = [8 8 8];
```

**Output:** swsub-XX\_pet\_avg.nii gzipped to sub-XX\_pet\_to\_MNI\_smoothed.nii.gz

## 14 Step 3: Quality Control

**Duration:** ~1 minute per subject

### 14.1 Setup and Load FSL

```
ml fsl/6.0.7.8

Define paths
BASE_DIR=~/Desktop/derivatives
SUB_DIR="${BASE_DIR}/data/sub-${subject_id}"
QC_DIR="${BASE_DIR}/QC/sub-${subject_id}"
mkdir -p "$QC_DIR"

Define image paths
T1="${SUB_DIR}/anat/sub-${subject_id}_T1w.reoriented.nii"
PET_MNI="${SUB_DIR}/pet/sub-${subject_id}_pet_to_MNI_smoothed.nii.
gz"
AVG_PET="${SUB_DIR}/pet/sub-${subject_id}_pet_avg.nii"
MNI_TEMPLATE="${FSLDIR}/data/standard/MNI152_T1_2mm_brain.nii.gz"

Centiloid masks
MASK_CEREB="data/references/centiloid_masks/voi_whlCbl_2mm.nii"
MASK_CTX="data/references/centiloid_masks/voi_ctx_2mm.nii"
```

### 14.2 Coregistration Check

Overlay averaged PET onto native T1 MRI to verify PET-MRI alignment:

```
fsleyes render --outfile "${QC_DIR}/qc_coreg.png" \
 --size 1200 400 \
 --scene ortho \
 "$T1" --overlayType volume --name "T1" \
 "$AVG_PET" --overlayType volume --name "PET" --cmap hot --
 alpha 50
```

**Expected result:** PET uptake should align with brain anatomy in MRI.

### 14.3 Normalization Check

Overlay MNI-normalized PET onto MNI152 template to verify spatial normalization:

```
fsleyes render --outfile "${QC_DIR}/qc_norm.png" \
 --size 1200 400 \
 --scene ortho \
 "$MNI_TEMPLATE" --overlayType volume --name "MNI152" \
 "$PET_MNI" --overlayType volume --name "PET_MNI" --cmap hot --
 alpha 50
```

**Expected result:** Brain boundaries should match MNI template.

### 14.4 Mask Alignment Check

Overlay Centiloid masks onto normalized PET to verify ROI placement:

```
fsleyes render --outfile "${QC_DIR}/qc_masks.png" \
 --size 1200 400 \
 --scene ortho \
 "$PET_MNI" --overlayType volume --name "PET_MNI" --cmap gray \
 "$MASK_CEREB" --overlayType volume --name "Cerebellum" --cmap
 Blue --alpha 40 \
 "$MASK_CTX" --overlayType volume --name "Cortex" --cmap Red --
 alpha 40
```

**Expected result:**

- Blue (Cerebellum) mask should cover cerebellar region
- Red (Cortex) mask should cover cortical gray matter

### 14.5 Review Checklist



A	B
Check	Pass Criteria
Coregistration	PET aligns with MRI brain boundaries
Normalization	Brain shape matches MNI template
Mask alignment	Cerebellum mask in posterior fossa, cortex mask on gray matter

**Expected Outputs:**

- QC/sub-XX/qc\_coreg.png -Coregistration overlay
- QC/sub-XX/qc\_norm.png-Normalization overlay
- QC/sub-XX/qc\_masks.png-Mask alignment overlay

**15 Step 4: SUVR and Centiloid Calculation****Duration:** ~2minutes per subject**15.1 Load Required Software**

```
ml fsl/6.0.7.8
ml freesurfer/7.3.2
```

**15.2 Define Centiloid Parameters**

For [11C]PiB with whole cerebellum reference:

```
CENTILOID_SUVR_ZERO=1.009 # Young control mean SUVR
CENTILOID_SUVR_100=2.076 # Typical AD mean SUVR
```

### 15.3 Resample PET to Mask Space

Ensure PET and masks have identical geometry:

```
flirt -in sub-XX_pet_to_MNI_smoothed.nii.gz \
 -ref voi_whlCbl_2mm.nii \
 -applyxfm -usesqform \
 -out sub-XX_pet_resampled_to_mask.nii.gz
```

### 15.4 Extract Regional Mean Values

```
Reference region (Whole Cerebellum)
ref_mean=$(fslstats sub-XX_pet_resampled.nii.gz -k
voi_whlCbl_2mm.nii -M)

Target region (Global Cortex)
target_mean=$(fslstats sub-XX_pet_resampled.nii.gz -k
voi_ctx_2mm.nii -M)
```

### 15.5 Calculate SUVR

```
suvr=$(echo "scale=6; $target_mean / $ref_mean" | bc -l)
```

### 15.6 Convert to Centiloid Scale

Using the linear transformation formula:

$$\text{Centiloid} = 100 \times (\text{SUVR} - \text{SUVR\_YC}) / (\text{SUVR\_AD} - \text{SUVR\_YC})$$

```
centiloid=$(echo "scale=4; 100 * ($suvr - 1.009) / (2.076 -
1.009)" | bc -l)
```

### 15.7 Save Results

Append to CSV file:



```
echo "$subject,$ref_mean,$target_mean,$suvr,$centiloid_value" >>
"$output_csv"
```

## 16 **Step 5: Statistical Validation**

**Duration:** ~1 minute

### 16.1 **Required Python Packages**

```
pip install pandas matplotlib scipy numpy
```

### 16.2 **Statistical Validation Script**

Save as statistical\_test.py:

```
#!/usr/bin/env python3
import pandas as pd
import matplotlib.pyplot as plt
import scipy.stats as stats
import numpy as np
import os
import sys

def main():
 # --- 1. CONFIGURE PATHS ---
 script_dir = os.path.dirname(os.path.abspath(__file__))
 project_root = os.path.dirname(os.path.dirname(script_dir))

 # Input files
 if len(sys.argv) > 1:
 results_file = sys.argv[1]
 else:
 results_file = os.path.join(project_root, "results",
 "tables", "all_subjects_results.csv")

 ref_file = os.path.join(project_root, "data", "references",
 "centiloid_values.csv")
 output_dir = os.path.join(project_root, "results", "reports")

 os.makedirs(output_dir, exist_ok=True)

 # --- 2. LOAD DATA ---
 print(f"Reading results from: {results_file}")
 df_calc = pd.read_csv(results_file)

 print(f"Reading reference from: {ref_file}")
 df_ref = pd.read_csv(ref_file)
 df_ref.columns = [c.strip() for c in df_ref.columns]

 print("Calculated Data Preview:")
 print(df_calc.head())

 # --- 3. MERGE DATASETS ---
 df_calc['subject_id'] =
df_calc['subject_id'].astype(str).str.strip()
 df_ref['Subject'] = df_ref['Subject'].astype(str).str.strip()

 df_merged = pd.merge(df_calc, df_ref, left_on='subject_id',
 right_on='Subject', how='inner')
```

```
if df_merged.empty:
 print("ERROR: No matching subjects found.")
 sys.exit(1)

print(f"Successfully merged {len(df_merged)} subjects.")

--- 4. CORRELATION ANALYSIS ---
SUVR Correlation
x_suvr = df_merged['global_cortical_suvr']
y_suvr = df_merged['SUVR']
r_suvr, p_suvr = stats.pearsonr(x_suvr, y_suvr)

Centiloid Correlation
x_cl = df_merged['global_cortical_centiloid']
y_cl = df_merged['Centiloid']
r_cl, p_cl = stats.pearsonr(x_cl, y_cl)

--- 5. PRINT RESULTS ---
print("\n=== STATISTICAL ANALYSIS RESULTS ===")
print(f"Number of Subjects: {len(df_merged)}")
print(f"\n1. SUVR Correlation:")
print(f" Pearson r: {r_suvr:.4f}")
print(f" p-value: {p_suvr:.4e}")
print(f"\n2. Centiloid Correlation:")
print(f" Pearson r: {r_cl:.4f}")
print(f" p-value: {p_cl:.4e}")
print("=====\n")

--- 6. GENERATE PLOTS ---
fig, axes = plt.subplots(1, 2, figsize=(12, 5))

Plot 1: SUVR
axes[0].scatter(x_suvr, y_suvr, alpha=0.7)
axes[0].set_title(f'SUVR Correlation\nr={r_suvr:.3f}, p={p_suvr:.3e}')
axes[0].set_xlabel('Calculated SUVR')
axes[0].set_ylabel('Reference SUVR')
m, b = np.polyfit(x_suvr, y_suvr, 1)
axes[0].plot(x_suvr, m*x_suvr + b, color='red', linestyle='--')

axes[0].grid(True, linestyle=':', alpha=0.6)

Plot 2: Centiloid
axes[1].scatter(x_cl, y_cl, color='green', alpha=0.7)
```

```
axes[1].set_title(f'Centiloid Correlation\nr={r_cl:.3f}, p={p_cl:.3e}')
axes[1].set_xlabel('Calculated Centiloid')
axes[1].set_ylabel('Reference Centiloid')
m, b = np.polyfit(x_cl, y_cl, 1)
axes[1].plot(x_cl, m*x_cl + b, color='red', linestyle='--')
axes[1].grid(True, linestyle=':', alpha=0.6)

plt.tight_layout()
output_plot = os.path.join(output_dir,
"correlation_plots.png")
plt.savefig(output_plot)
print(f"Plots saved to: {output_plot}")

if __name__ == "__main__":
 main()
```

### 16.3 Run Statistical Validation

python3 statistical\_test.py results/tables/all\_subjects\_results.csv

### 16.4 Expected Output

```
=== STATISTICAL ANALYSIS RESULTS ===
Number of Subjects: 25

1. SUVR Correlation:
 Pearson r: 0.9700
 p-value: 1.23e-15

2. Centiloid Correlation:
 Pearson r: 0.9700
 p-value: 1.23e-15
```

## Batch Processing

### 17 Create Batch Script

save as 00\_batch\_master.sh

```
#!/bin/bash
Master Batch Processing Script for PET-NeuroProject
set -e

Define subject range (default: 1-25)
start_sub=${1:-1}
end_sub=${2:-25}

Determine paths
SCRIPT_DIR="$(cd "$(dirname "${BASH_SOURCE[0]}")" && pwd)"
PROJECT_ROOT="$(dirname "$(dirname "$SCRIPT_DIR")")"

Output files
log_file="batch_processing_log_sub${start_sub}-${end_sub}_$(date +%Y%m%d_%H%M%S).txt"
master_csv="${PROJECT_ROOT}/results/tables/all_subjects_results.csv"

mkdir -p "$(dirname "$master_csv")"

Reset master CSV if starting from subject 1
if ["$start_sub" -eq 1] && [-f "$master_csv"]; then
 rm "$master_csv"
fi

echo "Starting Batch Processing: subjects $start_sub to $end_sub" | tee -a "$log_file"

for ((i=start_sub; i<=end_sub; i++)); do
 subject_id=$(printf "%02d" $i)

 echo "=====" | tee -a "$log_file"
 echo "Processing Subject: $subject_id" | tee -a "$log_file"
 echo "=====" | tee -a "$log_file"

 # Step 1: Data Organization
 echo "Running 01_data_org.sh..." | tee -a "$log_file"
 "${SCRIPT_DIR}/01_data_org.sh" "$subject_id" >> "$log_file"
 2>&1
 if [$? -ne 0]; then
```

```
 echo "ERROR: Data organization failed for $subject_id" |
tee -a "$log_file"
 continue
 fi

 # Step 2: Preprocessing
 echo "Running 03_preprocessing.sh..." | tee -a "$log_file"
 "${SCRIPT_DIR}/03_preprocessing.sh" "$subject_id" >>
"$log_file" 2>&1
 if [$? -ne 0]; then
 echo "ERROR: Preprocessing failed for $subject_id" | tee -
a "$log_file"
 continue
 fi

 # Step 3: Visual QC
 echo "Running visual_qc.sh..." | tee -a "$log_file"
 "${SCRIPT_DIR}/../qc/visual_qc.sh" "$subject_id" >>
"$log_file" 2>&1

 # Step 4: Analysis
 echo "Running analysis.sh..." | tee -a "$log_file"
 "${SCRIPT_DIR}/../Analysis/analysis.sh" "$subject_id"
"$master_csv" >> "$log_file" 2>&1

 echo "Finished Subject $subject_id at $(date)" | tee -a
"$log_file"
done

Generate QC Report
echo "Generating HTML QC Report..." | tee -a "$log_file"
python3 "${SCRIPT_DIR}/../qc/generate_report.py" >> "$log_file"
2>&1

echo "Batch Processing Complete. Results: $master_csv"
```

## 17.1 Running the Batch Script

```
Process all subjects (01-25)
./00_batch_master.sh 1 25

Process specific range
./00_batch_master.sh 5 10 # Only subjects 05-10

Resume from a specific subject
./00_batch_master.sh 15 25 # Subjects 15-25
```

## 17.2 Batch Script Workflow

```
For each subject (01 to 25):
├─ 01_data_org.sh → Extract & convert DICOM to NIfTI
├─ 03_preprocessing.sh → Motion correction, normalization,
smoothing
├─ visual_qc.sh → Generate QC images
└─ analysis.sh → Calculate SUVR & Centiloid

After all subjects:
└─ generate_report.py → Aggregate QC report
```

## 17.3 Monitoring Progress

```
Watch log file in real-time
tail -f batch_processing_log_*.txt

Check disk usage during processing
df -h
```

## 17.4 Expected Results

### Quality Metrics

- **Correlation (r):**  $\geq 0.95$  between calculated and reference values
- **Processing time:** ~20-30 minutes per subject (batch mode)

## 18 Troubleshooting



### Common issues

1. **Coregistration failure:** Check image orientations with fsorient -getorient
2. **Low correlation values:** Review QC images for specific subjects
3. **Missing frames:** Verify frame range in framing\_info.csv
4. **Disk space errors:** Enable cleanup in preprocessing script

### 19 Note

**FSL Pipeline:** Results from processing subjects 5, 9, 10 and 13 were outliers

**Hybrid FSL/SPM Pipeline:** Results from subjects 5, 9, and 10 were outliers

Outliers were excluded from the analysis.

### 20 Support

#### For questions or issues:

- GitHub Issues: <https://github.com/HolaDav/FLEX-PROJECT-PET-TEAM-O/issues>
- Email: oladejidavid12@email.com  
adeniyiabdullahi30@gmail.com

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