AMYQ: An index to standardize quantitative amyloid load across PET tracers

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Abstract

Introduction: Positron emission tomography (PET) amyloid quantification methods require magnetic resonance imaging (MRI) for spatial registration and a priori reference region to scale the images. Furthermore, different tracers have distinct thresholds for positivity. We propose the AMYQ index, a new measure of amyloid burden, to overcome these limitations.

Methods: We selected 18F-amyloid scans from ADNI and Australian Imaging, Biomarker & Lifestyle Flagship Study of Ageing (AIBL) with the corresponding T1-MRI. A subset also had neuropathological data. PET images were normalized, and the AMYQ was calculated based on an adaptive template. We compared AMYQ with the Centiloid scale on clinical and neuropathological diagnostic performance.

Results: AMYQ was related with amyloid neuropathological burden and had excellent diagnostic performance to discriminate controls from patients with Alzheimer’s disease (AD) (area under the curve [AUC] = 0.86). AMYQ had a high agreement with the Centiloid scale (intraclass correlation coefficient [ICC] = 0.88) and AUC between 0.94 and 0.99 to discriminate PET positivity when using different Centiloid cutoffs.

Discussion: AMYQ was related with amyloid neuropathological burden and had excellent diagnostic performance to discriminate controls from patients with Alzheimer’s disease (AD) (area under the curve [AUC] = 0.86). AMYQ had a high agreement with the Centiloid scale (intraclass correlation coefficient [ICC] = 0.88) and AUC between 0.94 and 0.99 to discriminate PET positivity when using different Centiloid cutoffs.

KEYWORDS
Alzheimer’s disease, amyloid burden standardization, amyloid pet, amyloid pet quantification, centiloid, neuropathology

1 | INTRODUCTION

Amyloid beta (Aβ) is one of the pathophysiological markers of Alzheimer’s disease (AD).¹² Amyloid imaging with positron emission tomography (PET) enables the in vivo evaluation of Aβ deposition in the brain.³-⁵ Several radiotracers have been developed for this purpose.¹¹C-Pittsburgh compound B (PiB), the first tracer, has been used widely since its inception.⁵ However, due to the short half-life of carbon 11, several fluorine-18 (¹⁸F)–labeled radioligands were developed: ¹⁸F-AV45, ¹⁸F-Florbetaben, or ¹⁸F-Flutemetamol. These PET tracers...
have proved to be a reliable alternative with wider application, and have therefore been accepted by the US Food and Drug Administration (FDA) and the European Medicines agency (EMA).4,6,7

Quantification of amyloid deposition in the brain is generally performed using a standardized uptake value ratio (SUVr).8 This methodology is used widely in both cross-sectional and longitudinal amyloid PET studies.9-12 However, it has some requirements and presents several caveats. First, this procedure depends on two regions of interest, one target cortical region in which to quantify amyloid and one reference region that should not be susceptible to amyloid pathology, used to intensity scale the PET images.13 The choice of the reference region is a source of variability, yielding different SUVr depending on which is used.13,14 Second, due to the low anatomical information of the amyloid PET images, the SUVr methodology typically requires an individual 3D magnetic resonance image (MRI) for spatial normalization and segmentation.15 Finally, the different amyloid tracers are not directly comparable and have different thresholds for amyloid positivity.15,16

To overcome these limitations, the Centiloid Working Group was created with the aim of scaling all non-standard methodologies of amyloid PET quantification into a standard Centiloid scale.17 This scale has shown good results across the different amyloid tracers.16,18-22 The Centiloid scale has been tested against the neuropathological results to determine amyloid positivity thresholds.19,23,24 However, the transformation of the SUVr values into Centiloid requires a structural MRI, a reference region, and a set of linear transformations.

Given that structural MRI is not systematically available, and that MRI motion artifacts could influence the accuracy of region of interest (ROI) definition,25 MRI-independent approaches have been developed recently to quantify amyloid deposition based on a spatial normalization process that does not require MRI.26,27 Some authors have proposed automated adaptive template methods, where a combination of different images is used to create a single template to mitigate the bias when a mean-single template is used.28-33 However, the methodologies developed thus far do not obtain standard amyloid measures across different tracers.

Our objectives were to develop a methodology that would be able: (1) to normalize amyloid PET data without the use of MRI, and (2) to quantify amyloid deposition in a reproducible manner across tracers (AMYQ), that does not require the use of a priori reference regions, or transformations between tracers. We assessed the relationship of AMYQ with standard neuropathological scales, and we compared it to the commonly used Centiloid scale.

2 | METHODS

2.1 | Participants

We selected a total of 751 subjects with available 18F-AV45 PET scans and 225 subjects with available 18F-Florbetaben PET scans from the Alzheimer’s Disease Neuroimaging Initiative (ADNI) cohort and 198 subjects from the Australian Imaging, Biomarker & Lifestyle Flagship Study of Ageing (AIBL) data set with available 18F-Flutemetamol PET scans. Participants were cognitive normal, had mild cognitive impairment (MCI), or an AD dementia diagnosis. All participants had a 3T T1-weighted MRI obtained <1 year apart from the PET acquisition. More details about acquisition and pre-processing steps on PET and MRI data can be found at adni.loni.usc.edu and aibl.csiro.au.

2.2 | Neuropathology assessment

Twenty-six subjects with 18F-AV45 PET had also a neuropathological assessment in ADNI (the mean interval between PET acquisition and death was 0.93 + 0.26 years). Pathological data included several amyloid quantification scales: the Consortium to Establish a Registry for Alzheimer’s Disease (CERAD) score,34 the Thal phase pathologic criteria,35 and the Alzheimer’s disease neuropathic change (ADNC).36 The CERAD score reflects the amyloid plaque density, particularly neuritic plaques in selected cortical areas, whereas the Thal phases reflect the topography of the amyloid plaques. This last assessment does not distinguish between compact and diffuse amyloid deposits. The ADNC is a composite score that uses the CERAD score and the Thal and Braak neurofibrillary tangle stages to classify the ADNCs. More information can be found at adni.loni.usc.edu.

2.3 | Image processing with MRI

All PET images were co-registered to their corresponding MRI using a rigid body registration and warped into the Montreal Neurological Institute (MNI) space using an MRI-based affine transformation followed by a non-linear registration using the Advanced Normalization Tools (ANTs) software.37 SUVs were calculated from the spatially normalized images using the defined cortical volume of interest as a composite and the whole cerebellum plus brainstem as the reference region.14,17 Finally, SUVs for the different tracers were transformed.
2.4 Template creation

To normalize the amyloid PET scans to the MNI regardless of MRI, we created a PET template in the MNI space. For each tracer, we iteratively and randomly selected 50 subsamples of MNI-normalized and scaled PET images of 15 controls, 10 MCI, and 15 AD subjects. For each subsample, we applied a principal component analysis (PCA) to obtain the main components. The final PCA template for each tracer was created as the mean of the 50 permutations of the previously obtained components. For all three tracers, the first two components of the PCA explained >90% of the variance of the data (Figure 1A). The visual inspection of each component identified the first component (PC0) as the non-specific binding of the amyloid tracer (ie, mainly white matter binding), whereas the second one (PC1) represented the specific cortical binding. More details about template creation are provided in the Supplementary Material.

2.5 Image processing with adaptive template and AMYQ index

All 1174 individual amyloid PET scans were normalized to the MNI space using a linear combination of the two components of the PCA template. Figure 1B shows the complete process. The native PET image was initially normalized with an affine registration to the PC0 using mutual information as similarity measure. To refine the normalization to MNI space, we performed an optimization process. First, we generated a subject-specific adaptive template optimizing the two betas of the two principal components of the PCA template ($\beta_0$ for the non-specific component, and $\beta_1$ for the cortical component) by maximizing the global correlation between the voxel intensities of the spatially normalized amyloid PET image and the adaptive template. Therefore,

$$\text{PET}_{\text{MNI-normalized}} = \beta_0 \cdot \text{PC0} + \beta_1 \cdot \text{PC1}$$

We then repeated the normalization step using the generated adaptive template using non-linear transformation from ANTs. Then we recomputed the betas to generate a new adaptive template with which to again normalize the images. This process was iteratively conducted until the betas were optimized and did not change from the previous normalization step. The algorithm finally created a unique amyloid-PET template for each subject using the optimal estimated betas for each of the two components of the template.

We assessed the normalization procedure of each tracer with the corresponding tracer-specific template (ie, data-driven specific and unspecific binding), but we also used the different tracer-specific templates in the other tracer subsamples. We compared the global Pearson’s correlation between templates, and they showed high correspondence between their specific and unspecific binding components (Supplementary Material). The $^{18}$F-Flutemetamol template outperformed the two other templates (ie, higher agreement with the Centiloid scale) in all subsamples. Therefore, we present the results using the PCA template created using $^{18}$F-Flutemetamol for all the analyses. For completeness, the results using the two other two templates are shown in the Supplementary Material.

The AMYQ index was determined as the $\beta_1/\beta_0$ ratio. We used the first component (or PC0) as a data-driven reference region to scale the cortical component and the second (or PC1) as the data-driven region.
in which to measure amyloid deposition. The ratio was then transformed into a 100-point scale using the same approach as that used by Klunk et al.\textsuperscript{17} Taking advantage of the $^{18}$F amyloid data set from Global Alzheimer’s Association Interactive Network (GAAIN),\textsuperscript{20–22} we calculated the mean $\beta_1/\beta_0$ ratio in the young controls and in the AD patients. The $\beta_1/\beta_0$ ratio of the young controls was mean centered to 0 AMYQ and to 100 AMYQ in AD patients. Thus the AMYQ for each individual was defined as:

$$\text{AMYQ} = 100 \times \frac{\beta_1/\beta_0 \text{Ind} - \beta_1/\beta_0 \text{YC}}{\beta_1/\beta_0 \text{AD} - \beta_1/\beta_0 \text{YC}}$$

As with the Centiloid scale, higher values of AMYQ represent high amyloid burden (ie, AD subjects mean scaled to 100 AMYQ), whereas values around zero represent low amyloid burden (ie, young controls mean scaled to zero). AMYQ, as the Centiloid scale, can exceed these limits.

### 2.6 Statistical analysis

Analyses were performed with the R statistical software (v 3.6.3; \url{https://www.r-project.org}). To assess the quality of the spatial normalization of the adaptive template methodology, we computed the similarity structural image similarity (SSIM)$^3$ index between the two normalized images. To warp the PET images using MRI and the PCA template was accurate across tracers (SSIM\textsubscript{AV45} = 0.85 + −0.06, SSIM\textsubscript{Florbetaben} = 0.85 + −0.05, and SSIM\textsubscript{Flutemetamol} = 0.91 + −0.04.

### 3 RESULTS

#### 3.1 Participants

PET images were successfully normalized with the adaptive template in 1148 subjects (of 1174) in contrast to the 1106 scans (of 1174), which where correctly warped to the MNI with the MRI-based normalization. Of note, 12 subjects failed with both methodologies; 56 failed only with the MRI-based normalization, whereas only 14 failed particularly with the adaptive template approach. Details on the normalization failure are provided in the Supplementary Material. Therefore, the normalization process was successful with both methodologies in 1092 subjects. Table 1 summarizes the demographics of these 1092 participants included in the subsequent analyses: 434 (39.7%) were cognitive normal, 505 (46.3%) had MCI, and 153 (14.0%) had an AD dementia diagnosis.

#### 3.2 Assessing quality control of normalization procedure

The estimated similarity between the spatially normalized PET images warped using MRI and the PCA template was accurate across tracers (SSIM\textsubscript{AV45} = 0.85 + −0.06, SSIM\textsubscript{Florbetaben} = 0.85 + −0.05, and SSIM\textsubscript{Flutemetamol} = 0.91 + −0.04.

#### 3.3 AMYQ correlates with the different neuropathological scales

Figure 2 shows the distribution of AMYQ and Centiloid scores across the different post-mortem amyloid quantifications. Both measures increased significantly with higher scores in the four neuropathologic scales ($\rho_{\text{AMYQ}} = 0.474$, $\rho_{\text{Centiloid}} = 0.562$; $\rho_{\text{AMYQ}} = 0.443$, $\rho_{\text{Centiloid}} = 0.551$; $\rho_{\text{AMYQ}} = 0.392$, $\rho_{\text{Centiloid}} = 0.546$; and $\rho_{\text{AMYQ}} = 0.610$, $\rho_{\text{Centiloid}} = 0.660$ for Thal phases, ADNC, and CERAD neuritic and diffuse, respectively). AMYQ and the Centiloid scale showed similar results in all the amyloid neuropathological stages. There were no statistically significant differences between the Thal phase 0 and phase 1, “none” versus “low” ADNC, or “none” versus “sparse” amyloid pathology using CERAD. Both AMYQ and Centiloid showed differences with the other Thal phases, when compared with Thal phase 0 or “none” pathology (P = .025 and P = .022 vs phase 4; P = .025 and P = .013 vs phase 5); with the ADNC scores (P = .012 and P = .010 vs “intermediate”; P = .024 and P = .015 vs “high”); and for both CERAD neuritic “frequent” (P = .029 and P = .005), and diffuse “frequent” plaques (P = .020 and P = .014 for AMYQ and Centiloid, respectively). The Wilcoxon ranked test did not reveal significant differences between AMYQ and the Centiloid scales in any of the scores of the four neuropathological scales.
TABLE 1  Demographics for the overall sample and the three subsamples divided by diagnosis (Control, MCI, mild cognitive impairment, and AD, Alzheimer’s disease dementia). Median and interquartile range (IQR) are shown for each continuous variable.

| Overall | AV45 | Florbetaben | Flutemetamol |
|---------|------|-------------|--------------|
|         | Control | MCI | AD | Control | MCI | AD | Control | MCI | AD | Control | MCI | AD |
| N | 434 | 505 | 153 | 150 | 405 | 121 | 144 | 61 | 15 | 140 | 39 | 17 |
| Age (years) | 71.60 [67.50, 76.00] | 72.00 [66.20, 77.00] | 75.00 [71.00, 79.70] | 72.80 [66.00, 76.20] | 75.30 [71.00, 79.50] | 69.70 [67.00, 74.53] | 74.00 [66.90, 78.30] | 73.60 [70.80, 79.70] | 72.00 [68.00, 76.00] | 74.00 [69.00, 78.50] | 75.00 [71.00, 79.00] |
| Gender = M (%) | 186 (42.9) | 283 (56.0) | 86 (56.2) | 73 (48.7) | 225 (55.6) | 68 (56.2) | 56 (38.9) | 38 (62.3) | 9 (60.0) | 57 (40.7) | 20 (51.3) | 9 (52.9) |
| Centiloid (median [IQR]) | 6.68 [−3.98, 26.77] | 73.47 [−0.16, 83.20] | 94.67 [55.37, 122.18] | 26.53 [−7.51, 32.25] | 28.28 [0.27, 83.20] | 94.28 [66.35, 121.74] | 6.70 [−7.51, 32.25] | 28.28 [0.27, 83.20] | 94.28 [66.35, 121.74] | 0.99 [−6.63, 15.74] | 6.70 [−7.51, 32.25] | 94.28 [66.35, 121.74] |
| AMYQ (median [IQR]) | 1.06 [−0.20, 27.30] | 90.29 [61.86, 111.16] | 90.29 [61.86, 111.16] | 22.38 [−10.17, 79.04] | 90.29 [61.86, 111.16] | 90.29 [61.86, 111.16] | 22.38 [−10.17, 79.04] | 90.29 [61.86, 111.16] | 90.29 [61.86, 111.16] | 22.38 [−10.17, 79.04] | 90.29 [61.86, 111.16] | 90.29 [61.86, 111.16] |

The AMYQ index had a high agreement with the Centiloid for the different PET tracers separately (ICCAV45 = 0.88, ICCFlorbetaben = 0.88, and ICCFlutemetamol = 0.89) and in the combined sample (ICC = 0.88). The Bland-Altman plot showed a bias of zero across the measurement range (r = −0.71, P < 0.001). When stratifying by amyloid positivity (as determined by a 12.2 Centiloid), the negative subsample showed a wider heterogeneity within the range values (r = 0.08; P > 0.05), whereas no bias was found in the positive subsample (r = −0.71; P < 0.001).
The discriminatory power of AMYQ to detect amyloid positivity was very high in the combined sample and in the three samples separately (AUCCombined = 0.94, AUCAV45 = 0.94, AUCAVFlorbetaben = 0.97, and AUCAVFlutemetamol = 0.89). An AMYQ threshold of 15.6 discriminated between amyloid positivity groups, with 92% specificity and 84% sensitivity in the combined sample. The aforementioned PET-positivity threshold of 15.6 AMYQ had a sensitivity of 86% and a specificity of 68% in the discrimination between controls and AD dementia patients in the overall sample. Of note, other established Centiloid thresholds to determine PET positivity23,39 yielded AUCs of AMYQ between 0.97 and 0.99 (Supplementary Material).

4 | DISCUSSION

The AMYQ index is a new metric to quantify brain amyloid load that does not require an MRI. AMYQ can be immediately computed in the nuclear medicine departments after the amyloid PET acquisition, providing a quantitative assessment that could help in clinical practice, which currently usually relies only on a visual read. Furthermore, AMYQ is interchangeable across tracers, and thus provides a standardized measure for amyloid quantification that could facilitate the comparison of studies in which different tracers are used and in clinical practice.

AMYQ is based on a synthetic amyloid template generated using PCA. The topographical pattern of this template is in agreement with previous tracer-specific works where one component represents the unspecific binding, mostly comprising white matter regions, and a second component that shows the neocortical uptake.32,33 Applying a linear combination of the specific and non-specific components by iteratively optimizing the weights of the two components of the PCA template, we created a unique adaptive template for each amyloid scan that will work as standard space to warp the individual image. By computing the ratio between these two weights, we obtained a measure of amyloid burden that accounts for the differences of intensity scale between PET acquisitions. One of the main advantages of this methodology is that it does not depend upon predefined flat (non-weighted) regions of cortical load or reference region to scale the PET. It is important to note that the weighted data-driven regions derived from the PCA analysis allow the assigning of more weight to key spatial regions related with the amyloid uptake pattern (eg, for the precuneus or medial frontal regions). Therefore, this approach removes the variability associated with the selection of distinct regions (both the reference region and the cortical region), and facilitates the harmonization between methods. To ensure the robustness of our results, we created a data-driven template for each tracer and compared the performance of each template onto all subsamples. All had good performances when applied to the other tracers, which is not
|                  | AV45 Florbetaben Flutemetamol | Centiloid AMYQ | Centiloid AMYQ | Centiloid AMYQ | Centiloid AMYQ |
|------------------|-------------------------------|----------------|----------------|----------------|----------------|
|                  | Hedges g | AUC    | Hedges g | AUC    | Hedges g | AUC    | Hedges g | AUC    | Hedges g | AUC    |
| CN—MCI           | 0.52     | 0.65   | 0.60    | 0.67    | 0.56    | 0.75    | 0.87    | 0.86    | 0.58    | 0.71    |
| MCI—AD           | 1.06     | 1.10   | 1.30    | 1.30    | 0.82    | 0.77    | 1.92    | 1.92    | 0.82    | 0.87    |

The discriminatory power of AMYQ when differentiating between AD dementia patients and controls showed excellent accuracy, equivalent to that of the Centiloid scale. For all three tracers, the AMYQ and the Centiloid scale index showed similar effect sizes and AUC to discriminate between the different diagnoses. In previous studies, Whittington et al.\(^3\) have shown a very good diagnostic performance of their amyloid load measure in \(^1\)\(^8\)F-AV45 samples. Their measure showed a higher sensitivity than the commonly used SUVr in all comparisons.\(^3\),\(^4\) Of note, in the \(^1\)^\(^8\)F-AV45 subsample of our work, AMYQ showed effect sizes similar to those reported by the Whittington measure. The agreement between the AMYQ and Centiloid values was excellent, both when assessed for the three tracers separately and when assessed in the combined sample. This yielded similar thresholds for AMYQ and the Centiloid scale when comparing controls and AD dementia patients. Of note, the thresholds were similar to those reported in other studies using the Centiloid scale.\(^5\) Of note, these thresholds were also similar to those reported when assessing agreement with CSF p-Tau/A\(_\beta\)\(_{42}\) and t-Tau/A\(_\beta\)\(_{42}\) ratios in the ADNI cohort.\(^3\)\(^9\) Despite the high correspondence between measures, we observed dissociation of values between the AMYQ and Centiloid in the subsample with low amyloid burden, but not in the amyloid-positive range.

The ability of AMYQ to detect amyloid positivity using different Centiloid cutoffs was also very high, with AUCs systematically above 0.94 in the overall sample and in the three subsamples separately. Using the Centiloid cutoff of 12.2 as the gold standard (this cutoff has been proposed to differentiate none/sparse from moderate/high CERAD scores and to identify A\(_\beta\)-detectable Thal phase),\(^2\)\(^4\) our analyses identified a very similar cutoff of 15.6 for AMYQ. When this threshold was used to differentiate patients with clinical diagnosis of AD...
dementia from controls, it proved very sensitive but with lower specificity for both scales; this low specificity is probably due to the high percentage of controls with amyloid pathology (40.8%). Higher Centiloid thresholds, between 23.5 and 26.0, might be better suited to identify intermediate/higher ADNC and, importantly, to correlate with visual reads.

Automated adaptive template methods have been used typically to normalize amyloid PET images, but not to replace the use of reference region to scale the intensity of the image. One of the most widely used methodologies to normalize amyloid PET without MRI is the CapAIBL,18,19,26 which reported excellent normalization results. Others have also reported an improved normalization using their respective adaptive templates in 18F-Flutemetamol scans32,46 or 11C-PiB.26

The advantage of AMYQ over the aforementioned automated adaptive template methods is that it provides a quantitative assessment of the amyloid load that does not require MRI and is interchangeable across tracers. Thus, AMYQ can be immediately computed in the nuclear medicine departments after the amyloid PET acquisition, and be used to complement the visual read. Visual reads have been classically implemented as a binary assessment. However, the number of borderline cases is non-negligible and, considering that the subthreshold ADβ positivity could be indicative of faster AD pathology accumulation, the dichotomization may overlook those at-risk individuals.47

Thus, AMYQ could aid in clinical practice as an immediate tool for PET quantification readily available at the nuclear medicine department. By using AMYQ, the practical difficulties imposed by the need of MRI usually acquired in a different visit in a different department or center, which have contributed to the limited implementation of quantification methods in clinical practice, are avoided.

Some limitations should be considered. First, this work proves that the AMYQ index is consistent using different templates randomly generated from a wide heterogeneous sample, but does not discard bias on homogenous samples. To determine whether the use of the “universal” Flutemetamol template is preferable over a single tracer template, we would require calculating AMYQ indices in a sample submitted with different radiotracers and PET scanners. Such samples were not available in this study. In addition, this work does not definitely solve whether the “universal” template should always be used instead of a specific template created for a particular study. The later approach might prove to be more accurate within a specific cohort, but it could compromise the comparability of the indices calculated. Second, the relationship between AMYQ and Centiloid in the amyloid-negative subjects should be further explored against a pathological gold standard. Moreover, this index has not been assessed in cohorts with other amyloid tracers such as 18F-NAV469 or 11C-PiB and it should be tested in other replication cohorts to validate the reported thresholds, and evaluated longitudinally to assess its sensitivity to change in time as this is a cross-sectional study. Despite the big sample size in some of the analysis, most of the sample had absence of neuropathological confirmation, which was in addition restricted to the 18F-AV45 subsample. Further analyses with larger sample size with neuropathological data should be assessed. Finally, Centiloid is widely used in research settings. While AMYQ could be quickly implemented in nuclear medicine departments, its use in research settings in which MRIs are usually acquired simultaneously will depend on its validation and proof of potential additional advantages, such as to measure longitudinal amyloid load or to capture early amyloid deposition.

In summary, our study provides a new index of global amyloid load that does not require a structural MRI and is independent from a priori reference regions. AMYQ could be used directly in clinical practice to quantify amyloid load consistently across amyloid tracers.

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COMPETING INTERESTS
AL has served at the scientific advisory boards of Biogen, Fujirebio Europe, Eli Lilly, Novartis, Roche Diagnostics, and Nutricia. JF has served at the scientific advisory boards of AC Immune and has provided consultancies to Novartis and Merck as part of the Medavante adjudication team in the Merck MK-8931 and Novartis CAP015A2201J and CCNP520A2202J trials. DA participated in advisory boards from Fujirebio-Europe and Roche Diagnostics and received speaker honoraria from Fujirebio-Europe, Nutricia, and from Krka Farmacéutica S.L. DA, JF, and AL declare a filed patent application (WO2019175379 A1 Markers of synaptopathy in neurodegenerative disease). All other authors report no competing interests.

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SUPPORTING INFORMATION
Additional supporting information may be found online in the Supporting Information section at the end of the article.

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