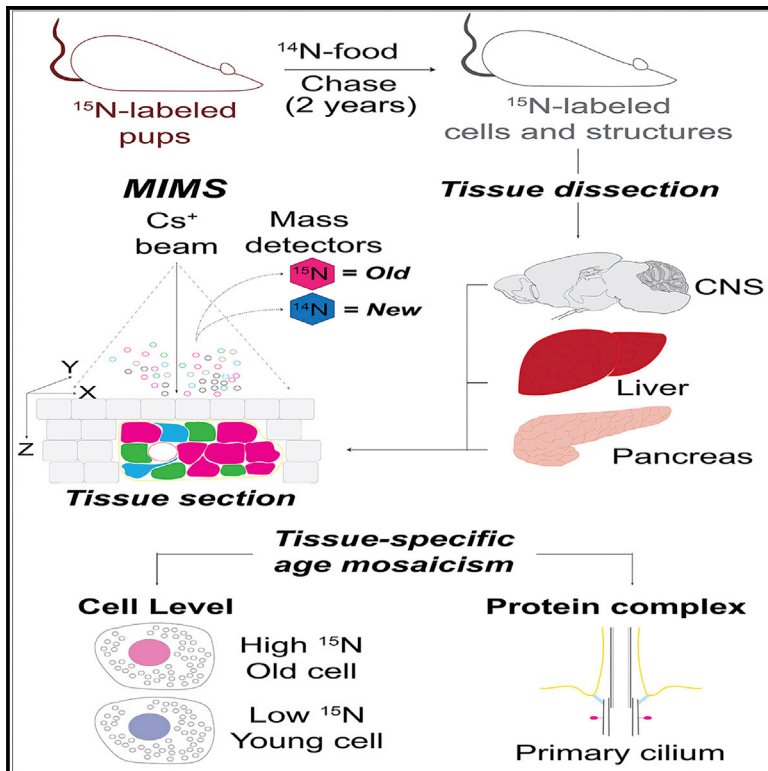


Cell Metabolism

Age Mosaicism across Multiple Scales in Adult Tissues

Graphical Abstract



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In Brief

Arrojo e Drigo et al. measure the age of cells and proteins using high-resolution isotope imaging and show that adult mouse organs are mosaics of cells of different ages. The liver, which has high turnover, contains cells as old as the animal, while cilia have differentially aged structural protein components.

Highlights

- Isotope microscopy reveals age mosaicism of cells and specific protein complexes
- Most hepatocytes are as old as neurons, while sinusoidal endothelial cells are young
- Pancreatic alpha, beta, and delta cells can be as old as the animal they reside in
- Components of the beta cell primary cilium last a lifetime



Age Mosaicism across Multiple Scales in Adult Tissues

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SUMMARY

Most neurons are not replaced during an animal's lifetime. This nondividing state is characterized by extreme longevity and age-dependent decline of key regulatory proteins. To study the lifespans of cells and proteins in adult tissues, we combined isotope labeling of mice with a hybrid imaging method (MIMS-EM). **Using ¹⁵N mapping, we show that liver and pancreas are composed of cells with vastly different ages, many as old as the animal.** Strikingly, we also found that a subset of fibroblasts and endothelial cells, both known for their replicative potential, are characterized by the absence of cell division during adulthood. In addition, we show that the primary cilia of beta cells and neurons contains different structural regions with vastly different lifespans. **Based on these results, we propose that age mosaicism across multiple scales is a fundamental principle of adult tissue, cell, and protein complex organization.**

INTRODUCTION

The lifespan of a terminally differentiated cell is quite variable among organs: **3 to 4 days for epithelial intestinal cells**, to olfactory neuronal cells replaced from basal stem cells, to a **life**

time for the majority of neurons, cardiomyocytes, and all inner ear hair cells (De Anda et al., 2016; Brann and Firestein, 2014; Foglia and Poss, 2016; Steinhäuser et al., 2012; Zhang et al., 2012). In some cases, somatic stem cells can respond to tissue damage and proliferate according to tissue-specific needs, as in the striated muscle, which can somewhat regenerate after wound because of activation of its satellite stem cells (Blau et al., 2015). A similar pattern is also observed at the proteome level, where proteins have different lifespans, ranging from hours and days to years (Ori et al., 2015; Toyama et al., 2013). **Like stressed and/or damaged cells, misfolded and damaged proteins must be degraded and replaced with new and functional versions** (Taylor and Dillin, 2011). Over the last decades these insights steered biomedical research toward focusing on cellular and molecular replacement processes.

It remains poorly understood how neurons, cardiomyocytes, and potentially other long-lived cells (LLCs), maintain functional integrity and protein homeostasis over the span of several decades. Because these cells are almost never replaced, they are essentially as old as the animal itself and must function properly throughout life, which in humans can be more than a century (De Anda et al., 2016). **Understanding how functionality is maintained in LLCs is important given that aging is associated with physiological impairments in these kinds of cells (e.g., neurons and cardiomyocytes)** (D'Angelo et al., 2009; Mattson and Magnus, 2006).

Recent advances in whole-animal metabolic-labeling strategies and quantitative mass spectrometry (MS/MS) have enabled system-wide, high-resolution analyses of global protein turnover

Context and Significance

Some cells in the body, such as skin, are constantly being replaced during one's lifetime, while most brain cells are not. Cells with minimal to no turnover contain long-lived proteins whose function declines with age. Researchers at the Salk Institute and University of California, San Diego, in California used a detailed isotope-tracing method to measure and map the age of cells and protein in different mouse organs. Unexpectedly, most organs are a mix of cells and proteins of vastly different ages, regardless of whether they are slow or quick to regenerate. Some cells also contain proteins of different ages. Understanding the principles of cellular aging will be important in helping combat the age-associated decline in organ function.



rates. However, imaging approaches for studying extreme cell longevity are not well developed. We recently combined stable isotope metabolic pulse-chase labeling of rats using ^{15}N (herein called ^{15}N -SILAM [stable isotope labeling in mammals]) with quantitative MS/MS to discover a class of long-lived proteins (LLPs) in neurons (Toyama et al., 2013). These neuronal LLPs localize primarily to the nuclear pore complex (NPC) and chromatin, and are maintained with no or limited turnover, in striking contrast to the majority of the proteome, which is renewed within hours or days (Ori et al., 2015; Schoenheimer, 1942). Only a few other LLPs have been previously identified, including lens crystalline, collagen, and myelin basic protein (Fischer and Morell, 1974; Lynnerup et al., 2008; Verzijlbergen et al., 2010). These proteins deteriorate with age and, with the exception of myelin, are unlikely to contribute to cellular aging because they typically reside in cells with minimal metabolic activity and/or play structural roles (Lynnerup et al., 2008). In contrast, the neuronal LLPs that we identified play critical roles in gene regulation and nuclear trafficking pathways and age-dependent loss of these LLPs impairs nuclear function (D'Angelo et al., 2009; Ibarra et al., 2016; Toyama et al., 2013).

LLCs face a constant lifelong demand for performance to maintain organ function and homeostasis and are constantly exposed to drivers of molecular and cellular damage. In fact, the accumulation of such damage over time may eventually alter gene expression and protein homeostasis pathways, leading to impaired cellular function and disease (D'Angelo et al., 2009; Ibarra et al., 2016). Understanding these mechanisms requires the identity and distribution pattern of LLCs in different organs. However, lifelong cell persistence has not been analyzed in a systematic and quantitative manner, and therefore fundamental questions about cell type-specific turnover rates remains poorly characterized. In addition, while implementation of ^{15}N -SILAM together with quantitative MS/MS allowed us to identify LLPs (Toyama et al., 2013), this approach lacks spatial information (other than the region of tissue that was dissected) and does not provide any information regarding cell identities. To overcome this challenge, we made use of ^{15}N -SILAM-labeled animals initially labeled for MS/MS to develop a new imaging pipeline for correlated scanning electron microscopy (SEM) and multi-isotope imaging mass spectrometry (MIMS) (herein called MIMS-electron microscopy [EM]). MIMS is a mapping technique compatible with electron microscopy that can determine the relative amount of different stable isotopes incorporated into nucleic acids and proteins in cells and tissues (Lechene et al., 2006; Steinhauser et al., 2012; Zhang et al., 2012). Given the high spatial resolution and sensitivity of EM and MIMS, we used MIMS-EM to visualize and quantify cell and protein turnover in the brain, liver, and pancreas in young and old ^{15}N -SILAM mice and rats.

Using MIMS-EM on tissue samples from exceptionally long-lived mice initially labeled for stable isotope mass spectrometry experiments (Toyama et al., 2013), we found that neurons and cardiomyocytes are not the only cell types with exceptional longevity. Much to our surprise, adult mouse liver and pancreas contain populations of cells with vastly different lifespans, many of which are as old as neurons. The pancreas serves a paradigm for the observed cellular age mosaicism, in that a subset of insulin-beta cells are as old as cortical neurons, while others

have been replaced through cell division during adulthood. In the liver, contrary to what has been inferred from liver regeneration models, we discovered that most hepatocytes, bile duct cells, and stellate-like cells (i.e., hepatic stellate cells and Kupffer cells) are LLCs. Age mosaicism can also be observed among capillary endothelial cells, which exhibit vastly different lifespans in different tissue such as the brain, pancreas, and liver. Furthermore, age mosaicism also occurs at the molecular level, where proteins with vastly different lifespans coexist at the supramolecular level in the basal body and axoneme of the primary cilium of neurons and beta cells.

RESULTS AND DISCUSSION

Identification of LLCs and Structures in the Central Nervous System (CNS) with MIMS-EM

Information about the lifespan of different cells is still lacking for many adult tissues, largely because of technical limitations. To determine which and whether cells divide or persist in a nondividing state throughout life we combined ^{15}N metabolic labeling of mice with MIMS-EM (Steinhauser et al., 2012). In brief, we used animals initially intended for protein MS/MS as in our previous studies (Toyama et al., 2013). As such, molecular components of these animals (i.e., nucleic and amino acids) were labeled with ^{15}N throughout their embryonic development and fed an ^{15}N -rich diet until the 21st or 45th day of post-natal development (see details in STAR Methods). We refer to these mice as P21- or P45- ^{15}N -SILAM mice, respectively. After the ^{15}N -labeling period, ^{15}N -animals were chased with a normal diet with the natural 99.4% content of ^{14}N for 6, 18, or 26 months of age. During the chase period, when ^{15}N -labeled cells are replaced by newly synthesized cells, ^{15}N is replaced by ^{14}N . Therefore, the rate of cell (and protein turnover) can be quantified with MIMS by simultaneously imaging ^{15}N and ^{14}N in a target region and determining the $^{15}\text{N}/^{14}\text{N}$ ratio (i.e., old-to-young ratio). Importantly, our previous work has shown that the relative amount of nuclear ^{15}N -labeled protein represents less than 2% of the whole nuclear proteome after 12 months of chase (Toyama et al., 2013). Therefore, the vast majority of the nuclear ^{15}N is expected to stem from ^{15}N -DNA and which decreases by 50% after each cell division, as expected (Steinhauser et al., 2012).

Since our SILAM approach cannot distinguish between ^{15}N -DNA and rare ^{15}N -proteins in the nucleus, we used MIMS-EM to quantitatively determine the ^{15}N content of random sections of neuronal nuclei in different regions of the CNS. This allowed us to establish a mean baseline and range of the nuclear ^{15}N signal that is characteristic for nondividing cells (De Anda et al., 2016; D'Angelo et al., 2009). We targeted neurons in the mouse layer-2 (L2) in the motor cortex and in the rat cerebellum from animals labeled with ^{15}N until P45 and chased for 6 months. MIMS-EM of the mouse L2 cortex showed that most of the ^{15}N signal was nuclear or overlapping with myelin sheaths, as expected (Figure S1C) (Toyama et al., 2013). We measured whole nuclei $^{15}\text{N}/^{14}\text{N}$ ratios and found that they were markedly higher than the natural occurrence ($^{15}\text{N}/^{14}\text{N} = 37 \times 10^4$) because of a significant retention of ^{15}N in all neuronal nuclei mapped (Figures 1A–1I and S1A–S1C). Since we observed species-specific differences (Figure S1E), we therefore decided to focus the rest of our analysis on mouse tissues.

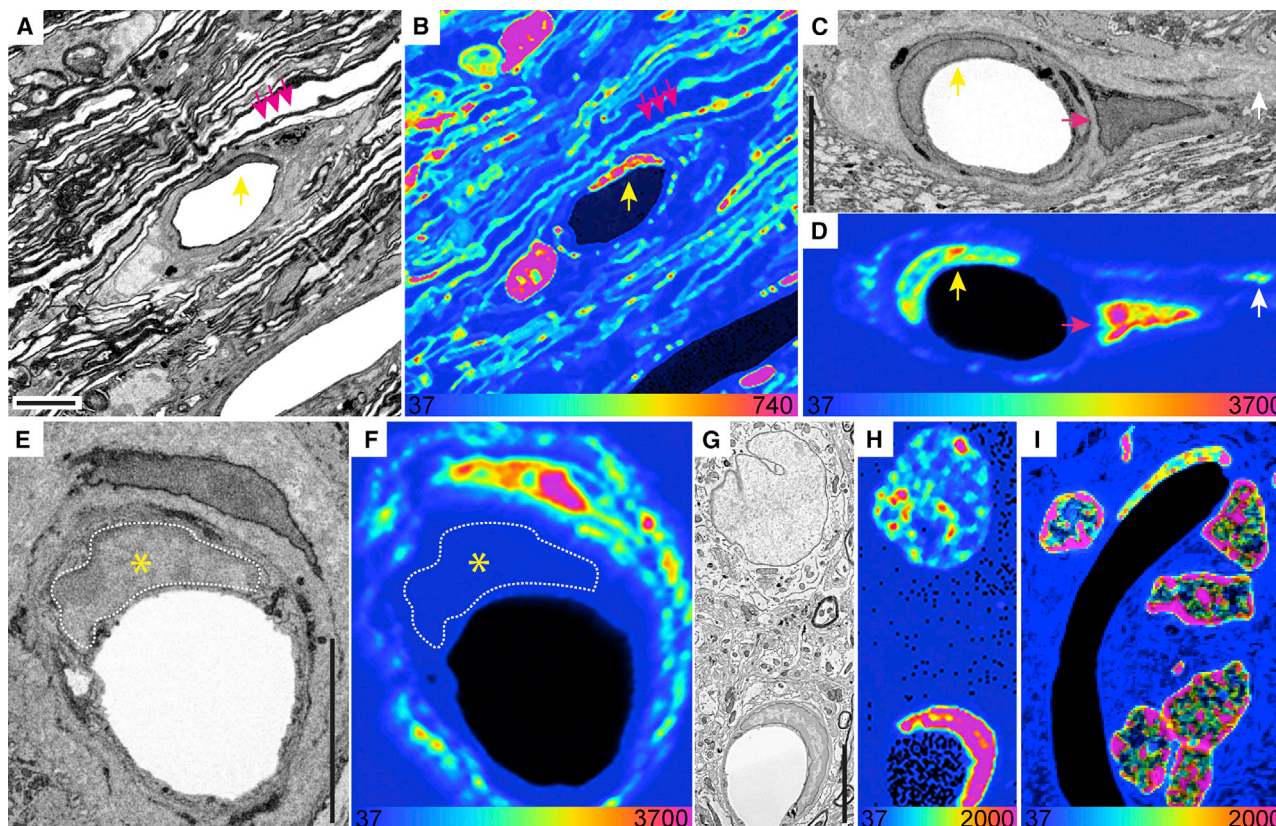


Figure 1. MIMS-EM of LLCs and Structures in the CNS

(A and B) SEM (A) and MIMS (B) of two capillaries in the optic nerve head (ONH) of a 6-month chase mouse. An endothelial cell nucleus (yellow arrow) and myelin sheaths (pink arrows) are indicated. A pericyte nucleus is visible to the left capillary lumen.

(C and D) SEM (C) and MIMS (D) of a capillary in the ONH. An old endothelial cell nucleus (yellow arrow), a fibroblast (pink arrow) and ^{15}N -rich extracellular matrix (ECM) (white arrow) are indicated.

(E and F) SEM (E) and MIMS (F) of a capillary in the ONH, with an old fibroblast (top) and a young endothelial cell (nucleus delineated in white and indicated by the yellow asterisk).

(G and H) SEM (G) and MIMS (H) of an L2 neuron (top) and an endothelial cell.

(I) Close-up of granular cells in the rat cerebellum with an old endothelial cell visible. Full mosaic is shown in Figure S1B.

Scale bars: 5 μm (A, C, E, and G) (SEM). At the bottom of the MIMS images, the heatmap shows the $^{15}\text{N}/^{14}\text{N} \times 10^4$ and scaled with a hue saturation intensity (HSI).

It is well established that cortical neurons do not divide during adulthood, thus we decided to use the mean $^{15}\text{N}/^{14}\text{N}$ ratio from random nuclear sections of L2 neurons as baseline level for the identification of nondividing cells. We observed that the mean $^{15}\text{N}/^{14}\text{N}$ in the L2 neuron population analyzed varied between 5- and 19-fold higher than the natural ratio (Figure S3D). This variation is likely explained by the random occurrence of dense heterochromatin patches in the nuclear sections imaged with MIMS-EM, whereby sections with a higher density of heterochromatin in a given section are expected to yield higher ^{15}N signals due to a relatively higher concentration of ^{15}N -DNA in comparison with euchromatin-rich sections. In fact, we observed a maximum level of ^{15}N labeling in intranuclear regions that correlate with dense patches of heterochromatin, which may also contain rare ^{15}N -proteins in addition to ^{15}N -DNA (Figures 1G, 1H, S1A, S1D, and S1E).

Accordingly, we decided to classify cells as “old” when their nuclear mean $^{15}\text{N}/^{14}\text{N}$ ratio was similar to L2 neurons,

namely $^{15}\text{N}/^{14}\text{N}$ ratios at least 5.5-fold over the natural ratio (i.e., $^{15}\text{N}/^{14}\text{N} > 208$, Figure S1F). Importantly, the relative age of non-neuronal cells in nervous and somatic organs was determined by comparing their $^{15}\text{N}/^{14}\text{N}$ with that of L2 neurons from P45- or P21-SILAM (e.g., cells from P45-SILAM animals were benchmarked against P45-SILAM L2 neurons). Of note, while this classification approach can determine the relative longevity of “old” cells after 6, 18, or 26 months of chase, it cannot determine their proliferative potential during the pulse period or how many times a cell has divided during the chase period (see below, Limitations of Study). Nevertheless, we found that cells in the inner nuclear layer of the retina; and most endothelial cells in the mouse optic nerve head (ONH) and all in the cerebellum were LLCs and as old as neurons, and did not show any sign of turnover (Figures 1A–1F and S1A–S1D). In fact, we observed retention of ^{15}N in endothelial cells located in the L2 cortical layer of a P21- ^{15}N -SILAM mouse chased for almost 26 months (Figures

1G and 1H). In addition, the nuclear $^{15}\text{N}/^{14}\text{N}$ ratio in perivascular fibroblasts and an oligodendrocyte was >20-fold higher than the natural $^{15}\text{N}/^{14}\text{N}$ ratio, similar to neurons (Figures 1A–1F and 4D). These data suggest that lifelong cell persistence extends to non-neuronal cells, is more prominent than previously thought and also confirms a recent report regarding oligodendrocyte longevity (Tripathi et al., 2017). In contrast, we found one example of a capillary endothelial cell in the ONH that had close-to-background ^{15}N levels (Figures 1E and 1F), which indicates that turnover of endothelial cells in the brain can happen, although it may be a rare event. Together, these data from these exceptionally long-lived SILAM mice provides the first evidence that the neurovascular unit, and, more specifically, endothelial cells in the brain, are mostly old. However, it also suggests that some vessels are mosaics, built of cells with vastly different lifespans (Figures 1A–1F). We conclude that MIMS-EM is a powerful tool to study cellular lifespan and we decided to explore the concept of age mosaicism in other tissues (see below).

Cellular Age Mosaicism in the Liver

Several studies in the context of tissue injury have led to the well-established concept that the liver has a high regenerative capacity (Malato et al., 2011). In line with this concept, our previous ^{15}N -SILAM MS/MS experiments in rats did not identify ^{15}N -labeled peptides in the liver after only a 4-month chase, which further suggested that these cells are replaced within months (Toyama et al., 2013). However, little is known about the longevity of hepatocytes and other key liver cell types in healthy adult mice (Magami et al., 2002; Malato et al., 2011). We applied MIMS-EM to map cells located near bile ducts, portal or central veins, and sinusoid capillaries (Figures 2A and 2B), and found that most hepatocytes in P45- ^{15}N -SILAM mice were as old as neurons. Long-lived hepatocytes were observed in the vicinity of central veins, portal veins, and capillary sinusoids after 6 and 18 months of chase (92% and 95%, respectively) (Figures 2A, 2B, and S2A–S2D). Similarly, MIMS-EM revealed that most cholangiocytes were old as neurons and, therefore, are also LLCs (red arrows in Figures 2A, 2B, 2D, 2F, and S2A–S2D). In contrast, while endothelial cells in portal and central veins did not divide (Figures 2A–2C, white arrows; Figures S2A–S2D), most sinusoid capillary endothelial cells contained ^{15}N levels at the 18-month time point that were lower than at the 6-month time point (Figures 2A, 2B, and 2E, white arrows; Figures S2A–S2D). A similar phenotype was observed in stellate-like cells (Figures 2A, 2B, and 2F, green arrows; Figures S2A–S2D). These results indicate that, while hepatocytes remain largely quiescent throughout life, endothelial cells in hepatic sinusoids and most stellate-like cells undergo a major turnover event between 6 and 18 months of life. Lastly, the exceptional longevity of liver cells was associated with old extracellular matrix elements located to the perivascular space of portal and central veins and interstitial space (Figures 2A–2D and S2B). These data suggest that the aging liver has a tissue architecture characterized by an intercellular age mosaicism, where a young sinusoid vasculature would support the function of relatively older cells, a notion that is further supported by hepatocyte lineage tracing studies that

estimate that hepatocytes could indeed last for at least 1 year (Magami et al., 2002; Malato et al., 2011).

Adult Pancreas is Composed of Cells with Vastly Different Lifespans

Having identified blood vessels containing endothelial cells with different ages, we asked whether age mosaicism is a more general principle and turned our attention to the pancreas, which contains exocrine and endocrine cells. Development and maturation of pancreatic endocrine and exocrine cells extends to the early postnatal period when cells are mostly generated by self-duplication (Dor et al., 2004; Houbracken and Bouwens, 2017). Here, we used MIMS-EM to determine the longevity and turnover of the endocrine and exocrine pancreas in adult animals labeled until two different time points in early life (post-natal day 21 [P21] or P45, shown in Figures 3A and 3B) and chased for almost 2 years. First, we analyzed islets from a P21- ^{15}N -SILAM mouse chased with ^{14}N for 26 months and observed that the turnover rate of alpha, beta, and delta cells is not uniform. As expected, we observed a large number of alpha and beta cells exhibiting background ^{15}N (representing 77% and 76%, respectively) (Figures 3F and S3D), likely reflecting the wave of alpha and beta cell proliferation that takes place in neonates and is important for beta cell maturation (Stolovich-Rain et al., 2015). Strikingly, almost a quarter of alpha and beta cells (representing 23% and 24%, respectively) and all delta cells had ^{15}N levels similar to neurons (Figure 3F), and therefore suggest that a subpopulation of islet endocrine cells can become LLCs before or at the weaning stage and may not proliferate for an entire lifetime.

Next, we analyzed islets from a P45- ^{15}N -SILAM mouse chased for 18 months and observed that most beta cells (61%) had ^{15}N levels approximately 10-fold higher than the natural ratio, similar to L2 neurons after 18 months of chase (Figure S3D), while 39% of beta cells had lower ^{15}N levels and therefore had proliferated during adulthood (Figures 3B–3G and S3D). Recent data indicate that different types of beta cells exist in the islet, including proliferation-prone and/or immature beta cells located in a neogenic niche at the islet periphery (Bader et al., 2016; van der Meulen et al., 2017). We mapped the location of proliferative beta cells in relation to the islet periphery and found no correlation between beta cell age and intraislet cellular location (Figure S3E), thus suggesting that beta cell proliferation is random within the islet ultrastructure. Similar results were observed for alpha cells, pancreatic endothelial cells, and stellate cells, which were also found to be mostly LLCs (Figures 3B–3G and S3D–S3F). In striking contrast, all delta cells and the majority of ductal cells did not show age mosaicism, and their ^{15}N levels were similar to neurons (Figures 3A–3G and S3B–S3D). In contrast, most acinar cells we observed had close to or background ^{15}N levels and thus have clearly undergone significant turnover during adulthood (Figures 3A and 3B, white arrow; Figure S3D), as expected (Houbracken and Bouwens, 2017). Notably, the observation of long-lived acinar cells suggests that some acinar cells become long-lived in early adult life (Figure 3G, pink arrow), which could have a role in pancreatic cancer (Pan et al., 2013; Westphalen et al., 2016; Wollny et al., 2016).

Together, these data indicate that the islet is a mosaic composed of cells with vastly different lifespans, where most alpha and beta cells assume a quiescent and LLC phenotype

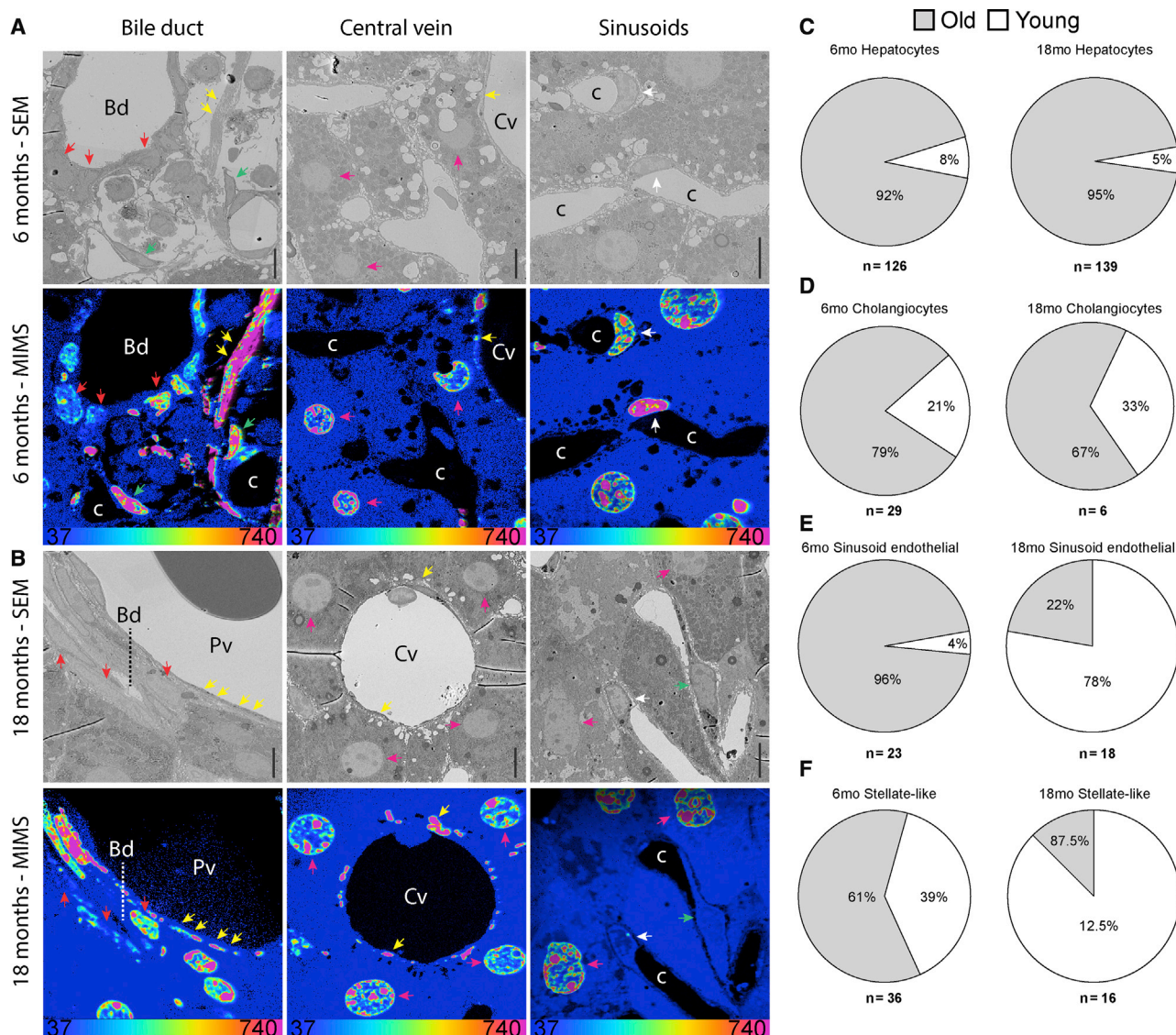


Figure 2. LLCs in the Liver

(A and B) SEM and MIMS of cells near bile ducts, central veins or capillary sinusoids from a ^{15}N -SILAM P45 mouse chased for 6 months (A) or 18 months (B). ECM (yellow arrowheads), hepatocytes (pink arrowheads), cholangiocytes (red arrowheads), stellate-like (green arrowheads), and endothelial cells (white arrowheads) are indicated. Cv, central vein; Bd, bile duct; c, capillary.

(C–F) Relative turnover of liver hepatocytes (C), cholangiocytes (D), sinusoid endothelial cells (E), and hepatic stellate-like cells (HSCs) (F) after a 6- or 18-month (mo) chase. The total number of cells analyzed for each cell type is indicated underneath each pie chart. At the bottom of the MIMS images, the heatmap shows the $^{15}\text{N}/^{14}\text{N} \times 10^4$ and scaled with a HSI. Scale bar, 10 μm .

between weaning (P21) and the first month (P45) of adult life. Moreover, it suggests that proliferation of alpha and beta cells is not uniform throughout adulthood, contrary to what earlier studies have indicated (Brennan et al., 2007; Dor et al., 2004). While the specific differences between young and old beta or alpha cells remains to be understood, our data adds to recent studies that have begun to investigate the cellular identity, the heterogeneity and aging of beta cell subpopulations and its potential role in age-associated type II diabetes onset (Aguayo-Mazzucato et al., 2017; Almaça et al., 2014; Basu et al., 2003; Johnston et al., 2016; van der Meulen et al., 2017; Segerstolpe et al., 2016).

Visualization of LLP Assemblies *In Vivo*

Our previous SILAM MS/MS experiments revealed that LLCs do contain a subset of proteins that persist throughout life. Specifically, we found that scaffold NPC proteins are maintained in the nuclear envelope throughout adulthood in neurons (Savas et al., 2012; Toyama et al., 2013). However, it was unclear whether this exceptional longevity extended to other subcellular components. While studying long-lived beta cells, we observed the retention of significant ^{15}N signal by a small, “star-shaped” structure within the cytoplasm of a beta cell from a P45- ^{15}N -SILAM mouse (data not shown). We hypothesized that this signal was associated with the basal body (Bb) of the cell’s primary

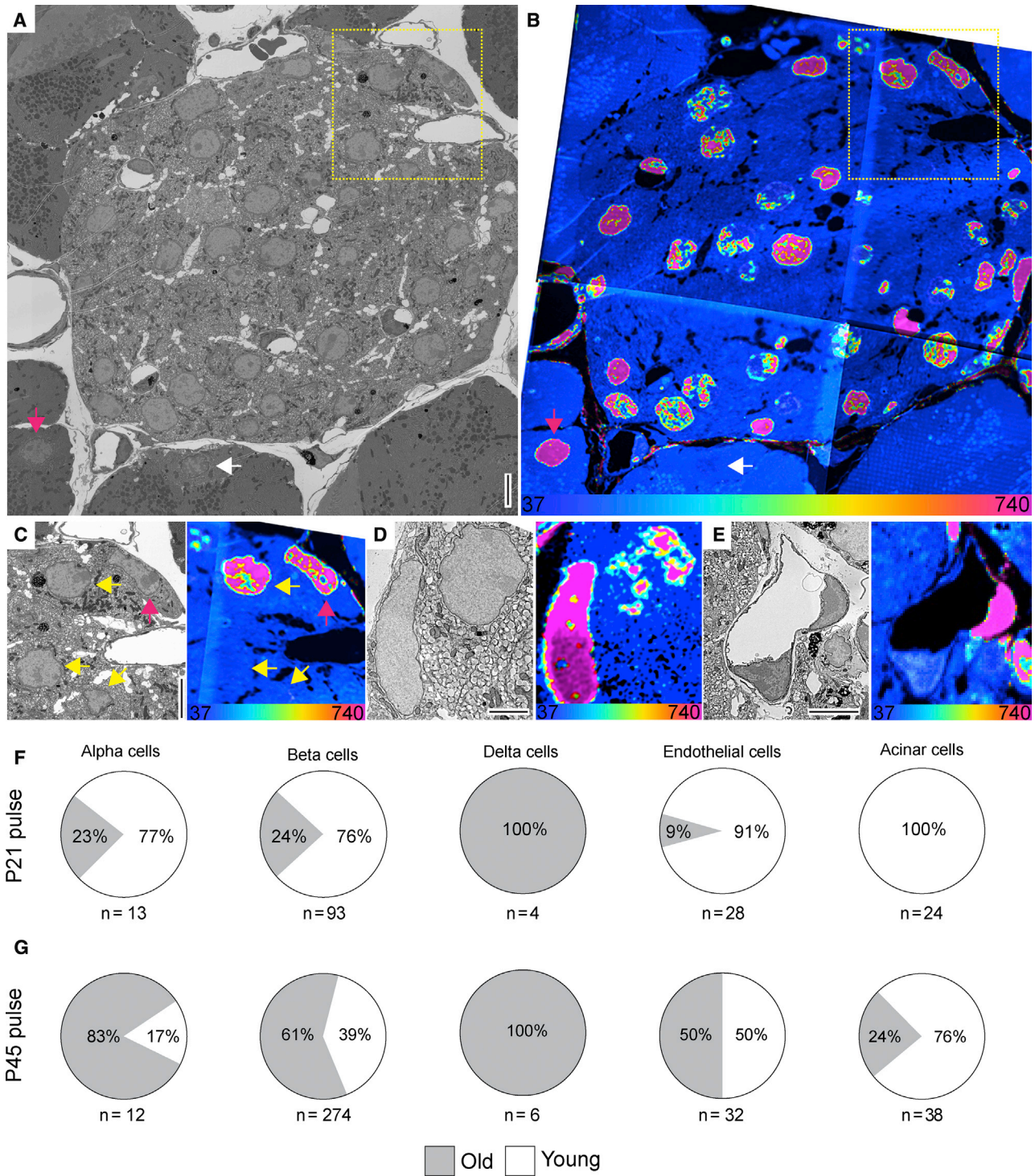


Figure 3. LLCs in the Pancreas

(A and B) SEM (A) and MIMS (B) of a cross-section of the islets of Langerhans. Old and young acinar cells are indicated by pink and white arrow, respectively. Yellow dotted box highlights cells shown in (C).

(C) Enlarged view of boxed region in (A) and (B). SEM and MIMS of beta cells (yellow arrows) and an old alpha cell (pink arrow).

(D) An old delta cell (left) next to a younger beta cell (top right).

(E) SEM and MIMS of a young (bottom) and an old (top right) endothelial cell. Old pancreatic stellate cells are seen in the top and lower right corners.

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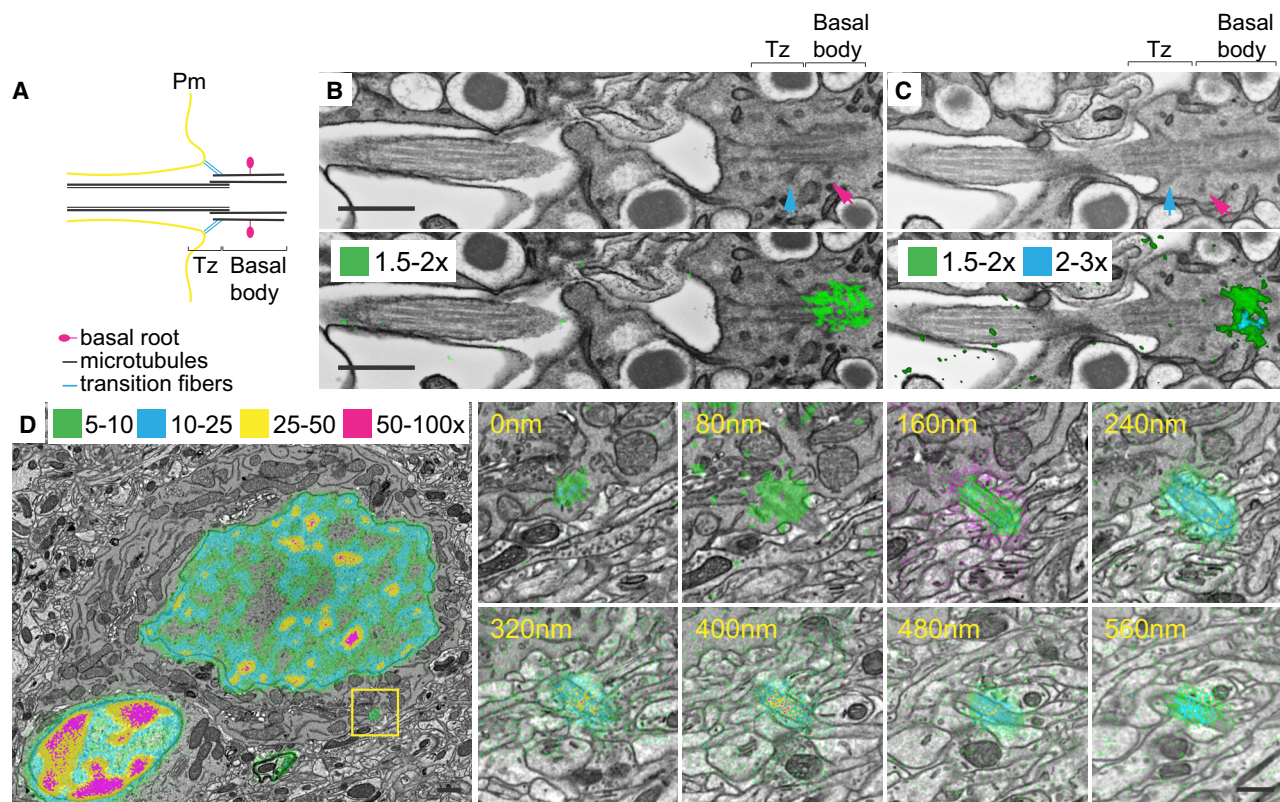


Figure 4. Longevity of Primary Cilia in LLCs

(A) Cartoon of a primary cilium Bb. Microtubules (black), transition fibers (cyan), basal root (pink), transition zone (Tz), axoneme, and plasma membrane (Pm) are shown.

(B and C) Two different sections of a beta cell primary cilium from a P21-¹⁵N-SILAM mouse chased for 26 months and imaged with MIMS-EM. Top rows: SEM micrographs with visible transition zone, Bb, transition fibers (cyan arrows), and basal root (pink arrows). Bottom rows: overlay of SEM micrographs and ¹⁵N/¹⁴N thresholds: 1.5×–2× (green) and 2×–3× (cyan) the natural ¹⁵N/¹⁴N of 37×10^4 .

(D) SEM micrograph overlaid with thresholded ¹⁵N/¹⁴N of an L2 neuron from a P45-¹⁵N-SILAM mouse chased for 6 months. The yellow box indicates the Bb. On the right are serial sections of the neuron Bb and primary cilium. ¹⁵N/¹⁴N thresholds: 5×–10× (green), 10×–25× (cyan), 25×–50× (yellow), and 50×–100× (magenta) the natural ¹⁵N/¹⁴N. At the bottom of the MIMS images, the heatmap shows the ¹⁵N/¹⁴N $\times 10^4$ and scaled with a HSI. Scale bars: 300 nm (B), 1 μ m (D), 500 nm (D, inset).

cilium (Figure 4A), which is a signaling organelle important for beta-cell function (Gerdes et al., 2014). We tested this hypothesis by imaging different beta-cell BBs from P21- and P45-¹⁵N-SILAM mice chased for 18 and 26 months, respectively. We applied MIMS-EM to serial cross-sections of different primary cilia within beta cells and discovered that, indeed, the Bb contained high ¹⁵N levels, while the rest of the cilium has been replaced by new components (Figures 4B, 4C, S4C, and S4D). We expanded this analysis to L2 neurons, where we observed that the ¹⁵N signal was not limited to the Bb and was also found in the axoneme, decreasing as the cilium projected toward the extracellular space (Figure 4D). These findings provide evidence of the occurrence of protein age mosaicism within one subcellular structure, while suggesting that long-lived structures are present in the primary cilium and that their turnover could be different in beta cells and neurons (Figures 4B–4D). While the identity of

these long-lived components remains to be discovered, it is noteworthy that long-lived nucleoporins, including Nup93, are known to form a permeability barrier in the ciliary gate in the cilium transition zone (McClure-Begley and Klymkowsky, 2017). These exciting findings raise the question of whether aging is associated with impaired primary cilium function due to loss of LLPs, as seen for NPCs in old neurons (D'Angelo et al., 2009).

Analogous to genetic mosaicism, which refers to the presence of two or more populations of cells with different genotypes in one individual, we propose the concept of age mosaicism as a key feature of adult tissue organization. Our data indicate that non-uniform turnover of cells and proteins gives rise to tissue- and cell-specific aging architectures characterized by a mosaic organization of young and old elements at the cell and protein level. These data provide insight into an organization pattern in biology that forms an age mosaic across the mesoscale, where the close

(F) Relative turnover in percentages of cells that are as old (gray) or younger (white) than L2 neurons from ¹⁵N-SILAM P21 mouse chased for 26 months.

(G) Same as in (F), but from a ¹⁵N-SILAM P45 mouse chased for 18 months. The total number of cells analyzed for each cell type is listed underneath each pie chart. At the bottom of the MIMS images, the heatmap shows the ¹⁵N/¹⁴N $\times 10^4$ and scaled with an HSI.

Scale bars: 5 μ m (A, C, and E) and 2.5 μ m (D).

relationship of young and old components is found in protein super complexes and organelles (i.e., Bb in the primary cilium), to cells (e.g., beta cells and hepatocytes) and tissues (e.g., brain, liver, and pancreas). **This study raises the exciting possibility that the age-dependent decline in LLPs is not limited to neurons, but may be a widespread phenomenon for all tissues rich in LLCs and links the loss of LLPs with the onset of age-related disease such as diabetes (Basu et al., 2003).** More broadly, future SILAM studies will contain the use of different stable isotope-labeled molecules (i.e., ^{15}N -thymidine for DNA and ^{18}O -Leucine for proteins) to help identify and functionally characterize other examples of heterogeneity in the aging of cells and proteins in different organs and in health and disease.

Limitations of Study

Our pulse-chase labeling approach using ^{15}N -enriched food can provide information about protein identity using MS/MS analyses (Toyama et al., 2013). However, our experimental set up using MIMS-EM cannot distinguish between ^{15}N -labeled nucleic acids and amino acids in the regions we investigated. Although most of the ^{15}N signal shown in brain, liver, and pancreatic cells is most likely due to ^{15}N -DNA, we cannot rule out a potential small contribution of ^{15}N -peptides from intranuclear LLPs (Toyama et al., 2013). However, regardless of the nature of the ^{15}N signal our results suggest that these cells are long-lived. Second, in its current stage, MIMS-EM is not a high-throughput imaging technique as it is time and labor intensive, requiring cells and tissues to be cut in thin sections (50–250 nm in thickness) and long acquisition times (e.g., 0.5–48 h for a $30 \times 30\text{-}\mu\text{m}$ raster), thus making volumetric imaging of any number of cells very challenging. As a result, this study was limited by the acquisition of random nuclear sections, which can add variability to the ^{15}N measurements by the random distribution of heterochromatin-rich domains in the sections analyzed. Nevertheless, we have imaged >700 cells in total with MIMS-EM to provide an important insight into the location and identity of LLCs outside the brain. MIMS-EM is a very sensitive and quantitative technique with multiplexed detection capabilities, and therefore future applications of MIMS-EM will take advantage of a SILAM approach that labels nucleic acids, amino acids, and fatty acids with different stable isotopes (^{15}N , ^{18}O , and ^{13}C , for example). Together with improved tools for cell/tissue mapping and data acquisition pipelines, MIMS-EM can be used to investigate a large number of cells and intracellular components from cells in virtually any SILAM tissue, including humans (Steinhauser et al., 2012).

STAR★METHODS

Detailed methods are provided in the online version of this paper and include the following:

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SUPPLEMENTAL INFORMATION

Supplemental Information can be found online at <https://doi.org/10.1016/j.cmet.2019.05.010>.

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AUTHOR CONTRIBUTIONS

R.A.eD. collected and analyzed the data and wrote the article. V.L.-E., S.T., R.R., T.D., E.B., S.P., and C.L. collected the data. V.O., M.H.E., and M.W.H. designed the study and wrote the article.

DECLARATION OF INTERESTS

The authors declare no competing interests.

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STAR★METHODS

KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Chemicals, Peptides, and Recombinant Proteins		
Cacodylate sodium salt (Cacodylate buffer pH 7.4)	Ted Pella Inc., Redding, CA	Cat no. 18851
Glutaraldehyde (2.5%)	Ted Pella Inc., Redding, CA	Cat no. 18426
Fresh Formaldehyde (2%)	Electron Microscopy Sciences, Hartfield, PA	Cat no. 18200
Calcium Chloride	Sigma-Aldrich	Cat no. 223506-500g
Potassium Ferrocyanide (3%)	J. T. Baker Chemical	Cat no. 3114
Osmium Tetroxide (4%, aqueous)	Electron Microscopy Sciences	Cat no. 19190
Thiocarbohydrazide	Electron Microscopy Sciences	Cat no. 21900
L-Aspartic Acid	Sigma-Aldrich	Cat no. A8949-25g
Lead Nitrate	Fisher Scientific	Cat no. L62-100
Ethanol (100 and 190 proof)	Koptec	Cat no. V1016 (100 proof) Cat no. V1105M (190 proof)
Durcupan ACM resin	Sigma-Aldrich, St Louis, MO	Cat no. 44611-500ml (Component A) Cat no. 44612-500ml (Component B) Cat no. 44613-100ml (Component C) Cat no. 44614-100ml (Component D)
Acetone	Fisher Chemical	Cat no. A949-4
¹⁵ N-mouse feed (u-15n, 98%+)	Cambridge Isotope Laboratories	Cat no. MF-SPIRULINA-N-S
Deposited Data		
Imaging data	Mendeley data	DOI: 10.17632/gkb5kz7vtp.1
Experimental Models: Organisms/Strains		
Mouse: FVB	The Jackson Laboratory, Bar Harbor, ME	Stock No: 001800
Software and Algorithms		
OpenMIMS	https://github.com/BWHCNI/OpenMIMS/	Steinhauser et al, 2012
Adobe Photoshop CS2015	Adobe	
Multi-color EM plug in for ImageJ	https://github.com/CRBS/multicolor_em/releases	Adams et al, 2016

CONTACT FOR REAGENT AND RESOURCE SHARING

For additional information and requests for reagents and resources, please contact the Lead Contact, Martin W. Hetzer (hetzer@salk.edu).

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Animal Studies

All mice used in this study were of the FVB/NJ strain (Stock No: 001800) and were acquired from the Jackson Laboratory (Maine, USA). All animal experimentation was approved by the University of California San Diego IACUC (protocol number S03172M).

METHOD DETAILS

Stable Isotope Metabolic Labeling of Mammals (SILAM)

Four FVB females were fed ¹⁵N-based spirulina chow (Cambridge isotope Laboratories, Inc.) starting at P45 for 10 weeks prior to introducing a male for 1 week. The females were kept on ¹⁵N chow diet during mating, gestation, and lactation. The Pups were weaned at P21. ¹⁵N-labeled pups were either kept on ¹⁵N chow up to P45 or weaned to ¹⁴N food after P21 (and then at P45). The pups were then euthanized, perfused with fixative as previously described ([Sosinsky et al., 2005](#)) at different time points representing

6, 18 or 26-months after the start of the ^{14}N chase. Briefly, mice were transcidentally perfused with 2.5% glutaraldehyde / 2% PFA in 0.15M sodium cacodylate containing 2mM calcium chloride and the brain, liver and pancreas were dissected and prepared for X-ray microscopy (XRM) and electron microscopy.

Processing of Samples for Electron Microscopy

After dissection, the tissue was cut into approximately 1mm^3 pieces and then postfixed in the same fixative at 4°C overnight. The specimens were washed with cacodylate buffer and then fixed in 2% osmium tetroxide / 1.5% potassium ferrocyanide in cacodylate for 1 hour at room temp. Specimens were thoroughly washed with water and then placed in 0.5% thiocarbonylhydrazide for 30 minutes and washed with water. The specimens were placed in 2% aq. osmium tetroxide for 1 hour and washed. The specimens were then placed in 2% aq. uranyl acetate at 4°C overnight. The specimens were washed with water and then placed into Walton's lead aspartate solution for 30 min at 60°C . The specimens were washed with water and then dehydrated on ice with 70% EtOH, 90% EtOH, 100% EtOH, 100% EtOH, and dry acetone (10 minutes each step). The specimens were placed into 1:3, 1:1, and 3:1 solutions of Durcupan ACM:acetone for 12 hours each. The specimens were placed into 3 changes of 100% Durcupan ACM for 1 day each and then placed into 60°C oven for 48 hours.

X-Ray Microscopy (XRM)

Identification and mapping of the 3D coordinates of islets in the exocrine pancreas was achieved by scanning epoxy-embedded pancreas blocks with X-ray microscopy (XRM, X-radia, Zeiss). Specimens were scanned with a Zeiss Versa 510 with the source set at 60 kV and 5W power using a 4X power objective (final pixel size ~ 2 microns depending on final geometric magnification). The specimens were rotated 360 degrees and 3201 projection images were collected. We were able to identify islets in situ due to their x-ray contrast differences with the surrounding tissue (Figure S2A, inset A' and Video S1). This allowed us to visualize and reconstruct large volumes of the mouse pancreas block in 3D and estimate the volume of islets in situ (Figures S1A and S1B). Next, based on the resulting microCT scan, we used a Leica ultramicrotome to trim the specimen block down to the targeted islet and 250-nm-thick sections were collected on silicon wafers for MIMS imaging.

Correlative Electron Microscopy and Multi-isotope Mass Spectroscopy (MIMS)

The MIMS technique is capable of measuring spatially localized concentrations of isotopes in the sample and hence, providing accurate chemical maps (Lechene et al., 2006; Steinhauser et al., 2012; Zhang et al., 2012). However, the XY spatial resolution in the MIMS is limited to about ~ 50 nm (Herrmann et al., 2007), in comparison the achievable resolution in the SEM is typically about a nm or less (Goldstein et al., 2003). To obtain 2D or 3D chemical information from our samples, single or serial sections of 50-250nm were cut and arranged on a Si wafer. Large field of view SEM micrographs of 80-250nm-thick sections of brain, liver and pancreas were acquired using a GeminiSEM (Zeiss, Germany) guided by automated tile acquisition using AtlasAT software (Fibics, Ottawa, Canada) and a pixel size of 4nm. Endogenous islets were imaged with XRM to acquire their 3D coordinates prior to guide islet sectioning using an ultra-microtome and prepared for SEM-AtlasAT mapping. MIMS image acquisition of mapped areas of interest and organelles was done as previously described (Steinhauser et al., 2012; Zhang et al., 2012). Briefly, brain, liver and pancreas tissue sections containing different mapped cells and organelles of interest were imaged with a NanoSIMS 50L (Cameca, France) using a cesium (Cs^-) beam and ^{15}N and ^{14}N levels were detected simultaneously.

MIMS-EM Image Analysis and Display

The chemical maps (i.e. $^{15}\text{N}/^{14}\text{N}$ ratiometric images) obtained by the Cameca NanoSIMS were overlaid on the SEM image of the same section, after alignment and post-processing. Therefore, in the overlay image, the chemical sensitivity of the MIMS image and the spatial resolution of the SEM image are both preserved. These post-processing steps used to create the overlay image are explained below. The SEM images of the serial sections were aligned using the affine transformations of the StackReg plug in of ImageJ (Thévenaz et al., 1998) or using a python-based alignment tool developed in house (NCMIR) to compensate for any image distortions due to sectioning. The MIMS images were resized to match the pixel size of the SEM image. Since the NanoSIMS is a destructive imaging technique, and it tends to have preferential sputtering rates depending on the atomic species being sputtered (Lechene et al., 2006; Steinhauser et al., 2012; Zhang et al., 2012), the images were considerably distorted with respect to the SEM images. Just using an affine transformation was not sufficient to get a good correlation between the MIMS and the SEM image, and images had to be adjusted for warping. The typical distortion map of the MIMS image when aligned to the SEM image is shown in Figures S4A and S4B. All the linear and warp transformations of the MIMS image with respect to the SEM image was performed using Photoshop CS2015 (Adobe, USA), and the transformations were recorded. The recorded transformations were then applied to the chemical map of interest, which in our case is the $^{15}\text{N}/^{14}\text{N}$ ratio map. The ratio maps were computed using the openMIMS plug in in ImageJ. The aligned $^{15}\text{N}/^{14}\text{N}$ ratio map was then thresholded to split it into 4 images, representing signal levels from low to high. A histogram stretching algorithm was applied to all the 4 images, so that signal was spread from 20000 to 63353 (maximum value of 16-bit images). The 4 images were color coded to green, cyan, yellow and magenta; with green being the lowest signal and magenta being the highest signal. To overlay the color coded $^{15}\text{N}/^{14}\text{N}$ ratio map on the SEM image, a transparency factor was added to the color, so that it did not obscure the gray scale information. An ImageJ plug in 'Multi Color EM' was developed for this purpose, and details of the algorithm used are available (Adams et al., 2016). This plug in is available online (https://github.com/CRBS/multicolor_em/releases).

QUANTIFICATION AND STATISTICAL ANALYSIS

Quantification of $^{15}\text{N}/^{14}\text{N}$ Ratio in Tissues and Cell Age Classification

Quantification of MIMS ^{15}N , ^{14}N and $^{15}\text{N}/^{14}\text{N}$ ratio data was done with the Fiji plug in OpenMIMS (available at <https://github.com/BWHCNI/OpenMIMS/>) (Steinhauser et al., 2012; Zhang et al., 2012). Briefly, we quantified the intensity of ^{15}N and ^{14}N in each pixel inside the nucleus of individual cells in the brain, liver and pancreas. Next, we determined the $^{15}\text{N}/^{14}\text{N}$ ratio for each pixel and generated the ratiometric images shown in the paper. To determine the relative age of cells in the brain, liver and pancreas, we used the average nuclear $^{15}\text{N}/^{14}\text{N}$ of L2 cortical neurons from old animals (e.g., 18- or 26-months of chase) from which the target tissue was dissected. This L2 $^{15}\text{N}/^{14}\text{N}$ was therefore used as a “ $^{15}\text{N}/^{14}\text{N}$ standard value” that would be expected for non-dividing LLCs (Figures S1–S3, n=60 for P45-18mo chase and n=44 for P21-26mo chase). Next, and based on the fact that cells lose ~50% of nuclear ^{15}N after each cell division (Steinhauser et al., 2012), we determined that young cells (i.e. younger than L2 neurons) would have an average nuclear $^{15}\text{N}/^{14}\text{N}$ value that was lower than the respective L2 $^{15}\text{N}/^{14}\text{N}$ mean value \pm standard deviation of the data (S.D.). Similarly, old cells (i.e. as old as neurons) would have a $^{15}\text{N}/^{14}\text{N}$ mean value that was at least equal or higher than the respective L2 $^{15}\text{N}/^{14}\text{N}$ mean value \pm S.D. The number (n) of each cell type quantified in liver and pancreas can be found under the pie charts in Figure 2 and Figure 3, respectively.

Cell Type Classification in Electron Micrographs

Cells were classified according the following anatomical hallmarks: in the CNS, neurons were identified as cells that had synaptic connection to the cell body and relatively electron “light” chromatin; glial cells were identified as cells with electron dense heterochromatin patches and next to neurons (see Figure 4); endothelial cells were identified by their often thin and elongated nuclei and presence of a capillary lumen. In the liver, hepatocytes were identified by their relatively large nuclei (sometimes cells were bi-nucleated) and the presence of electron-dense granular structures in the cytosol that correspond to glycogen stores, hepatic endothelial cells were found lining the sinusoid capillary (these were often found to be perforated), central or portal vein lumen. Stellate-like cells were identified by their elongated and fibroblast-like morphology and their cytosol usually contained lipofuscin bodies and the cells were in the tissue interstitium or near vessels. In the pancreas, beta cells were identified by the morphology of the insulin granules with condensed electron-dense crystals surrounded by a large halo; alpha cells were identified by the morphology of the glucagon granules with dark and electron-dense crystals surrounded by a small halo, delta cells by the morphology of the somatostatin granules with electron-light lumen and no halo, endothelial cells were lining the capillary lumen and were fenestrated. Pancreatic stellate cells were classified using the same parameters as hepatic “stellate-like” cells. Acinar cells were identified by their dense endoplasmic reticulum network and the presence of large secretory granules. Ductal cells were found lining the epithelium of pancreatic ducts and that were in the interstitium of the exocrine pancreas.

DATA AND SOFTWARE AVAILABILITY

All data and software used in this manuscript are available upon request to the Lead Contact (hetzer@salk.edu). The images shown in the main figures in this manuscript are available online (Mendeley Data DOI [10.17632/gkb5kz7vtp.1](https://doi.org/10.17632/gkb5kz7vtp.1)).

ADDITIONAL RESOURCES

Please visit <https://ncmir.ucsd.edu/sbem-protocol> for an online version of the protocol used to process samples for electron microscopy.