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A Cell- and Tissue-Specific Weakness of the Protein Homeostasis System Underlies Brain Vulnerability to Protein Aggregation

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HIGHLIGHTS

A branch of the protein homeostasis system regulates protein aggregation

This system is weaker in brain tissues than in other body tissues

This system is weaker in Braak regions than in other brain regions

This system is weaker in neurons than in other brain cell types

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A Cell- and Tissue-Specific Weakness of the Protein Homeostasis System Underlies Brain Vulnerability to Protein Aggregation

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SUMMARY
The phenomenon of protein misfolding and aggregation is associated with a wide range of neurodegenerative conditions that cause progressive loss of function in specific regions of the human brain. To understand the causes of the selective cell and tissue vulnerability to the formation of these deposits, we analyzed the ability of different cell and tissue types to respond, in the absence of disease, to the presence of high levels of aggregation-prone proteins. By performing a transcriptional analysis, we found that the protein homeostasis system that regulates protein aggregation is weaker in neurons than in other cell types and in brain tissues than in other body tissues. These results suggest that the intrinsic level of regulation of protein aggregation in the healthy state is correlated with the selective vulnerability of cells and tissues to protein misfolding diseases.

INTRODUCTION
Neurodegenerative disorders are multifactorial conditions characterized by extensive neuronal dysfunction associated with the misfolding and aggregation of a specific set of proteins that form aberrant deposits, including amyloid plaques and neurofibrillary tangles in Alzheimer’s disease and Lewy bodies in Parkinson’s disease (De Strooper and Karran, 2016; Eisenberg and Jucker, 2012; Holtzman et al., 2011; Selkoe and Hardy, 2016; Knowles et al., 2014). Increasing evidence also indicates that protein aggregation is a widespread phenomenon, as hundreds of proteins unrelated in sequence or structure have been found to form insoluble assemblies under conditions of stress, aging, or disease (Chapman et al., 2006; Ciryam et al., 2015, 2017; David et al., 2010; Gidalevitz et al., 2006; Tartaglia et al., 2007; Walther et al., 2015).

To rationalize these observations, it has recently been reported that a large fraction of the proteome is inherently metastable against aggregation within the cellular environment (Ciryam et al., 2013, 2015, 2016, 2017; Yerbury et al., 2019). Proteins are metastable in their native states when their cellular concentrations exceed their intrinsic solubilities (Tartaglia et al., 2007; Vecchi et al., 2020). Indeed, proteins that coaggregate in plaques, tangles, and Lewy bodies have been found to be highly metastable against aggregation (Ciryam et al., 2013), and biochemical pathways associated with neurodegenerative diseases, including in particular oxidative phosphorylation, have been reported to be enriched in these proteins, hereafter referred to as metastable proteins, thus revealing a common feature for these otherwise very different neurodegenerative conditions (Ciryam et al., 2013). Moreover, a fraction of these metastable proteins were reported to be specifically transcriptionally downregulated in Alzheimer’s disease (Ciryam et al., 2016). This metastable subproteome associated with Alzheimer’s disease was found to be associated primarily with specific components of the protein trafficking and clearance mechanisms, in particular the endosomal-lysosomal and the ubiquitin-proteasome systems (Kundra et al., 2017). These results suggest that these quality control systems are crucial for the maintenance of the cellular homeostasis of a set of metastable proteins prone to aggregation in Alzheimer’s disease.

In addition, the protein quality control mechanisms, which are collectively known as the protein homeostasis system, have been shown quite generally to become progressively impaired upon aging and in neurodegenerative diseases, resulting in increased accumulation of aggregated species (Balch et al., 2008; Freer et al., 2016; Freer et al., 2019; Fu et al., 2018, 2019; Hipp et al., 2014; Knowles et al., 2014). Hence, it is crucial to maintain a balance between the levels of aggregation-prone proteins and their associated protein homeostasis components in order to preserve overall cellular health. Further studying these effects may therefore help answer the question of why a large number of the diseases associated with protein aggregation involve the central nervous system. Although there have been a number of studies about the vulnerability of specific brain tissues (Bero...
et al., 2011; Mattson and Magnus, 2006; Miller et al., 2013), it is still not completely clear why the brain itself is more vulnerable to the aggregation phenomenon compared with other body tissues.

The aim of the present study is to address this question by investigating whether healthy brains are more susceptible to the effects of protein misfolding diseases than other organs, by examining at the transcriptional level the balance between the expression of aggregation-prone proteins and the efficacy of their corresponding protein homeostasis control system. To achieve this goal, we determined the expression levels of metastable proteins associated with Alzheimer’s disease and their associated protein homeostasis components across nearly 80 different healthy human tissues, including several ones from the brain. Our expectation is that brain tissues, compared with those from other regions of the body, should have a higher level of expression of the set of metastable proteins but a lower expression of the associated homeostasis components. In this view, due to the inherent lower expression levels of the quality control components associated with metastable proteins, certain regions of the brain would not have the ability to mount a robust response to these highly aggregation-prone metastable proteins, suggesting an origin for the selective vulnerability of the brain to protein misfolding diseases.

RESULTS
The Strength of the Protein Homeostasis System that Responds to Protein Aggregation Is Proportional to Level of Gene Expression of a Metastable Subproteome
Because aggregation-prone proteins are intrinsically metastable even in the absence of disease, we studied healthy brain tissues to identify their associated protein homeostasis components using an approach based on coexpression network analysis (see Supplemental Information). We used a subset of these metastable proteins specifically associated with Alzheimer’s disease (Ciryam et al., 2016), as our aim was to study the vulnerability of different tissues to this disease. In this way, using a weighted coexpression approach (Langfelder and Horvath, 2008), we identified the protein homeostasis components (PHC) associated with this metastable subproteome (MS) (Kundra et al., 2017). We then used here extensive microarray data obtained for 77 different types of healthy human tissues (Su et al., 2004) to study the vulnerability of various tissues based on the relative expression of these genes (Table S1). We carried out this analysis by defining a protection factor, s, as the correlation of the relative expression levels of the MS and its corresponding PHC. When s = 1 the response to protein aggregation in a given tissue matches exactly, at the transcript level, the requirement to prevent aggregation of the metastable proteins, whereas values below 1 indicate a lower, and possibly ineffective, response. We found from this analysis that the brain tissues have an elevated expression of genes corresponding to the MS, but not of their protein homeostasis counterpart, relative to other types of tissues in the body (Figure 1A). Although in body tissues the PHC expression level grows more rapidly than that of the MS (s = 1.33), in brain tissues the opposite is true (s = 0.88). Thus, although the protein homeostasis response in all tissues tends to be generally proportional to the pressure toward protein aggregation, brain tissues tend to behave differently from other tissues (Figure 1B) and have an overall weaker response.

Vulnerable Brain Tissues Have a Weaker Response to the Pressure toward Protein Aggregation Than Non-vulnerable Ones
The greater vulnerability of brain tissues to the pressure toward protein aggregation observed above suggests the presence of a link between the levels of regulation of protein aggregation and the development of neurodegenerative processes in Alzheimer disease. In order to investigate this relationship in more detail, we differentiated between tissues that are vulnerable and those that are resistant to neurodegeneration in Alzheimer’s disease, as assessed by the Braak staging (Braak and Braak, 1991). We mapped different types of brain tissue to the respective Braak stages as reported previously (Freer et al., 2016; Freer et al., 2019). Seven of these tissues corresponded to different Braak stages present in the microarray dataset used in the current study. We analyzed the levels of expression of the MS and the associated PHC in these types of tissue and compared them with the tissues not affected in Braak stages. We found that brain tissues corresponding to Braak regions have a lower value of s than do those of the non-Braak regions (Figure 2). These results reveal a close link between the regulation of the pressure toward protein aggregation with tissue vulnerability to Alzheimer’s disease.

Neurons Have a Lower Ability to Maintain Protein Homeostasis Than Other Types of Brain Cells
These results suggest a possible origin for the increased vulnerability of neurons to changes that lead to Alzheimer’s disease relative to other non-neuronal brain cell types. In order to verify whether or not neurons are less able
Figure 1. The Protein Homeostasis System that Responds to Protein Aggregation Is Weaker in Brain Tissues Compared with Other Tissues in the Human Body

(A) Comparison between the average expression of genes corresponding to the metastable subproteome (MS) and the average expression of the genes corresponding to the associated protein homeostasis components (PHC) in 77 different types of human tissue (Table S1). The protection factor $s$ defines the strength of the protein homeostasis response to the presence of metastable proteins. Brain tissues have smaller values of $s$ (orange, $s = 0.88$) than other tissues of the body (blue, $s = 1.13$), indicating that brain tissues have a reduced ability to protect themselves from protein aggregation relative to other tissues.

(B) Correlation matrix of gene expression levels in different tissues, where the correlation is calculated for the genes corresponding to the MS and their associated PHC. The intensity of the colors and the size of the circles are proportional to the correlation coefficients. Brain tissues (first row) are only weakly correlated with other body tissues, as also illustrated by the body image on the bottom left. The statistical significance was evaluated as explained in Methods and in Figure S1.
than other brain cell types to regulate protein aggregation, we compared the expression of genes corresponding to the MS and their associated protein homeostasis components (PHC) in neurons, astrocytes, microglia, and oligodendrocytes, using extensive single-cell RNA sequencing data (Habib et al., 2017; Lake et al., 2016). We found that neurons have the lowest value of the protection factor $s$ among these cell types (Figure 3), indicating that neurons are indeed the cell types most vulnerable to protein aggregation in the brain.

Proteins in the Oxidative Phosphorylation Pathway Have a Reduced Ability to Maintain Protein Homeostasis in the Brain Than in Other Tissues

In order to understand in more detail the molecular origins of the increased vulnerability of certain types of brain cells and tissues to the effects leading to Alzheimer's disease, we considered the ability of the cells to regulate oxidative phosphorylation, because this is a specific process associated with metastable proteins (Ciryam et al., 2016; Kundra et al., 2017), and more generally with neurodegenerative diseases (Lin and Beal, 2006; Rhein et al., 2009). We divided the genes associated with the oxidative phosphorylation pathway into two parts, those that are metastable and those that are not (Table S2). We then calculated the value of protection factor $s$ for the...
oxidative phosphorylation genes and their associated PHC for both groups across all tissues. The difference in $s$ values between the body and brain tissues ($\Delta s$) is higher for the group of oxidative phosphorylation genes that are metastable, whereas there is almost no difference in the values of $s$ for non-metastable oxidative phosphorylation genes. These results suggest that, even within the oxidative phosphorylation pathway, those genes that are also metastable pose a higher risk for the brain tissues, as these are less well prepared to face the challenge posed by the presence of these aggregation-prone proteins (Figure 4).

DISCUSSION

Taken together, the results of this study suggest that the strength of the protein homeostasis system that responds to protein aggregation is a central element in determining the vulnerability of cells and tissues to protein misfolding diseases. Our analysis shows that brain tissues are less effective than other body tissues in dealing with the problems posed by protein aggregation. The protection factor $s$ serves as an effective measure of the vulnerability of different tissue types to protein aggregation, as it decreases progressively from body tissues to resistant brain tissues to highly vulnerable brain tissues (Figures 1 and 2). It also provides insights into the molecular origins of Alzheimer’s disease, as neurons have the lowest value of $s$ of all the types of cells found in the central nervous system (Figure 3).

As the human brain experienced a rapid cortical reorganization during the course of recent evolution from non-human species, one may speculate that the expression levels of critical genes could also have increased too rapidly for the simultaneous evolution of the regulatory components of the protein

Figure 4. Proteins in the Oxidative Phosphorylation Pathway Have a Weaker Protein Homeostasis Response to Protein Aggregation in the Brain than in Other Tissues

(A) Average expression of the genes corresponding to the metastable subproteome (MS) within the KEGG oxidative phosphorylation (OP) pathway (inset) and their associated PHC.
(B) Average expression of all genes corresponding to non-metastable proteins within the OP pathway and their associated PHC. The value of $s$ for these critical genes is lower (<1) throughout the body, not just in the brain, and the difference in the values of $s$ between the body brain tissues is more pronounced for the genes encoding for metastable proteins in the oxidative phosphorylation pathway. The subunits in pink in the schematic show the location of the metastable genes within the oxidative phosphorylation pathway. The colour code is the same as in Figure 1.
homeostasis system. In this view, when faced with the challenges posed by aggregation-prone proteins, brain tissues are more likely to suffer from protein aggregation diseases.

In conclusion, our results indicate that the specific vulnerability of certain brain tissues to protein aggregation is associated with their relative inability to regulate this process already in the healthy state. Furthermore, we have also provided evidence that this inability is characteristic of neurons, which are therefore observed to be the earliest affected by protein misfolding diseases, compared with other brain cell types. These observations link the molecular origins of this neurodegenerative disease to the specific branch of the protein homeostasis network that regulates the maintenance in their soluble state of the proteins that are metastable against aggregation.

Limitations of the Study
In this study, we investigated the cell- and tissue-specific capacity of the protein homeostasis system to respond to the pressure caused by the overall expression of aggregation-prone proteins. Although useful insights can be derived from a transcriptome-level analysis, as we have done here, a proteome-level analysis will in future studies provide more stringent conclusions.

It is possible that specific genetic traits of different individuals could influence the results in the transcriptional analysis carried out using just six control healthy brains from the Allen Brain Atlas. We also note that the transcriptional analyses of brain and non-brain tissues were carried out on different datasets, again leaving open the possibility that genetic traits of specific individuals could influence the results.

METHODS
All methods can be found in the accompanying Transparent Methods supplemental file.

SUPPLEMENTAL INFORMATION
Supplemental Information can be found online at https://doi.org/10.1016/j.isci.2020.100934.

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AUTHORS CONTRIBUTIONS
All the authors participated in the design of the study, its execution, the analysis of the results, and the writing of the manuscript.

DECLARATION OF INTERESTS
We declare no conflict of interests.

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Supplemental Information

A Cell- and Tissue-Specific Weakness of the Protein Homeostasis System Underlies Brain Vulnerability to Protein Aggregation

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Supporting Information

Transparent Methods

Dataset Acquisition

**Healthy brain tissues.** Microarray data for healthy brain tissues were acquired from the Allen Brain Atlas (Hawrylycz et al., 2012). Gene expression data for 6 healthy human brains were available, across 900 different tissues. Data were scaled and normalised using the ‘scale’ function of the Weighted Gene Correlation Network (WGCNA) package (Langfelder and Horvath, 2008). As the Allan Brain Atlas uses multiple probes for each genes, the ‘collapseRows’ function of WGCNA package was used to get a single expression value for each gene across all samples. The expression values for each gene was then averaged across all six brains to arrive at the final expression value associated with each gene (Kundra et al., 2017).

**Tissue specific data.** We obtained tissue specific data across various human tissues from a previously published dataset (Su et al., 2004). Certain cancerous tissues and cell lines were removed from the analysis as our aim was to study the expression levels in a healthy state, leaving 77 tissues for the study (Table S1). From these, 27 were characterised as neural tissues and 50 body tissues (Table S1). For the Braak and Non-Braak analysis, brain regions (from the 27 brain tissues) were assigned to either Braak or Non-Braak (Freer et al., 2016). Briefly, brain regions in the Allen Brain Atlas were matched with the closest regions mentioned in the original paper (Braak and Braak, 1991). 7 tissues were found to correspond to different Braak stages (Table S1).

**Cell type specific data.** Single-cell RNAseq data for 4 different cell types found in the brain were obtained from a published dataset (Darmanis et al., 2015). Data were scaled and normalised in R.

**Coexpression network construction**

We used the data obtained from the Allen Brain Atlas to construct a coexpression network for the genes corresponding to the metastable subproteome associated with Alzheimer’s disease (MS) (Kundra et al., 2017) and the genes associated with the protein homeostasis system to get the protein
homeostasis complement (PHC) of the MS. WGCNA was used to construct the coexpression network (Kundra et al., 2017). Briefly, WGCNA is a clustering algorithm based on hierarchical clustering but which uses ‘soft thresholding’ and the concept of topological overlap or shared neighbours to identify clusters of coexpressed genes. The soft thresholding method assigns a weight to each pair of interacting genes and uses such weight along with the topological overlap to identify modules of coexpressed genes in the expression data.

**Calculation of the protection factor s**
Since our aim was to study the balance between the expression level of aggregation prone proteins and components associated with their regulation, we defined a protection factor, $s$, as the slope of the best-fit line for the relative expression of the MS and its PHC across different tissue or cell types. A line was fitted based on linear regression between the expression of MS and the PHC. The protection factor is a measure of the strength of the balance between the MS and its PHC.

**Statistical testing**
To evaluate the significance of our results, we used random sets of genes corresponding in number to the MS and PHC and calculated the $s$ for them. We then calculated the difference in $s$ between the body and brain tissues ($\Delta s$). We repeated these 1000 times to have a frequency distribution of $\Delta s$ values for random sets of genes. The $\Delta s$ for our genes of interest is 0.25, which is more than two SD away from the random sets of genes (Figure S1). Thus, our $s$ values are highly robust and significant. The statistical testing was performed using the Scipy module in Python. The comparison of different ‘$s$’ values was done using a two-tailed t test with a value of $\alpha$ of 0.001.
| Tissue                          | Classification (1) | Classification (2) |
|--------------------------------|--------------------|--------------------|
| Adipocyte                      | Body               |                    |
| Adrenal cortex                 | Body               |                    |
| Adrenal gland                  | Body               |                    |
| Appendix                       | Body               |                    |
| Atrioventricular Node          | Body               |                    |
| BDCA4+DendriticCells           | Body               |                    |
| Bone marrow                    | Body               |                    |
| Bronchial Epithelial Cells     | Body               |                    |
| CD105+_Endothelial             | Body               |                    |
| CD14+_Monocytes                | Body               |                    |
| CD19+_BCells(neg._sel.)        | Body               |                    |
| CD33+_Myeloid                  | Body               |                    |
| CD34+                          | Body               |                    |
| CD4+_Tcells                    | Body               |                    |
| CD56+_NKCells                  | Body               |                    |
| CD71+_EarlyErythroid           | Body               |                    |
| CD8+_Tcells                    | Body               |                    |
| Cardiac Myocytes               | Body               |                    |
| Fetal Thyroid                  | Body               |                    |
| Fetal liver                    | Body               |                    |
| Fetal lung                     | Body               |                    |
| Heart                          | Body               |                    |
| Kidney                         | Body               |                    |
| Liver                          | Body               |                    |
| Lung                           | Body               |                    |
| Lymphnode                      | Body               |                    |
| Ovary                          | Body               |                    |
| Pancreas                       | Body               |                    |
| Pancreatic Islet               | Body               |                    |
| Placenta                       | Body               |                    |
| Prostate                       | Body               |                    |
| Salivary gland                 | Body               |                    |
| Skeletal muscle                | Body               |                    |
| Skin                           | Body               |                    |
| Smooth muscle                  | Body               |                    |
| Testis                         | Body               |                    |
| Testis Germ Cell               | Body               |                    |
| Testis Interstitial            | Body               |                    |
| Testis Leydig Cell             | Body               |                    |
| Testis Seminiferous Tubule     | Body               |                    |
| Thymus                         | Body               |                    |
| Thyroid                        | Body               |                    |
| Tongue                         | Body               |                    |
| Tonsil                         | Body               |                    |
| Tissue                        | Location | Classification |
|------------------------------|----------|----------------|
| Trachea Body                 | Body     | -              |
| Uterus Body                  | Body     | -              |
| Uterus Corpus Body           | Body     | -              |
| Whole Blood Body             | Body     | -              |
| colon Body                   | Body     | -              |
| Small intestine Body         | Body     | -              |
| Amygdala Brain               | Brain    | Braak          |
| Caudate nucleus Brain        | Brain    | Non Braak      |
| Cerebellum Brain             | Brain    | Non Braak      |
| Cerebellum peduncles Brain   | Brain    | Non Braak      |
| Ciliary ganglion Brain       | Brain    | Non Braak      |
| Cingulate cortex Brain       | Brain    | Braak          |
| Dorsal root ganglion Brain   | Brain    | Non Braak      |
| Fetal brain Brain            | Brain    | Non Braak      |
| Globus pallidus Brain        | Brain    | Non Braak      |
| Hypothalamus Brain           | Brain    | Braak          |
| Medulla oblongata Brain      | Brain    | Non Braak      |
| Occipital lobe Brain         | Brain    | Braak          |
| Olfactory bulb Brain         | Brain    | Non Braak      |
| Parietal lobe Brain          | Brain    | Non Braak      |
| Pons Brain                   | Brain    | Non Braak      |
| Prefrontal cortex Brain      | Brain    | Braak          |
| Spinalcord Brain             | Brain    | Non Braak      |
| Subthalamic nucleus Brain    | Brain    | Non Braak      |
| Superior cervical ganglion   | Brain    | Non Braak      |
| Temporal lobe Brain          | Brain    | Braak          |
| Thalamus Brain               | Brain    | Braak          |
| Trigeminal ganglion Brain    | Brain    | Non Braak      |
| Whole brain Brain            | Brain    | Non Braak      |
| Pineal day Brain             | Brain    | Non Braak      |
| Pineal night Brain           | Brain    | Non Braak      |
| Pituitary Brain              | Brain    | Non Braak      |
| retina Brain                 | Brain    | Non Braak      |

Table S1. List of the 77 different tissues used in the analysis. The Braak classification (Braak and Braak, 1991) is reported for brain tissues, related to Figure 1.
Figure S1. Histogram showing $\Delta s$ (difference in $s$ for body tissues and brain tissues) for 1000 random sets of genes, related to Figure 1. The $\Delta s$ for our genes of interest (genes that are supersaturated and downregulated only in Alzheimer’s disease, and the associated PHC) is 0.25 (shown by red arrow).
Figure S2. Neurons have a lower value of protection factor $s$ compared to other non-neuronal cell types, related to Figure 3. *** $p<0.001$, see Methods.