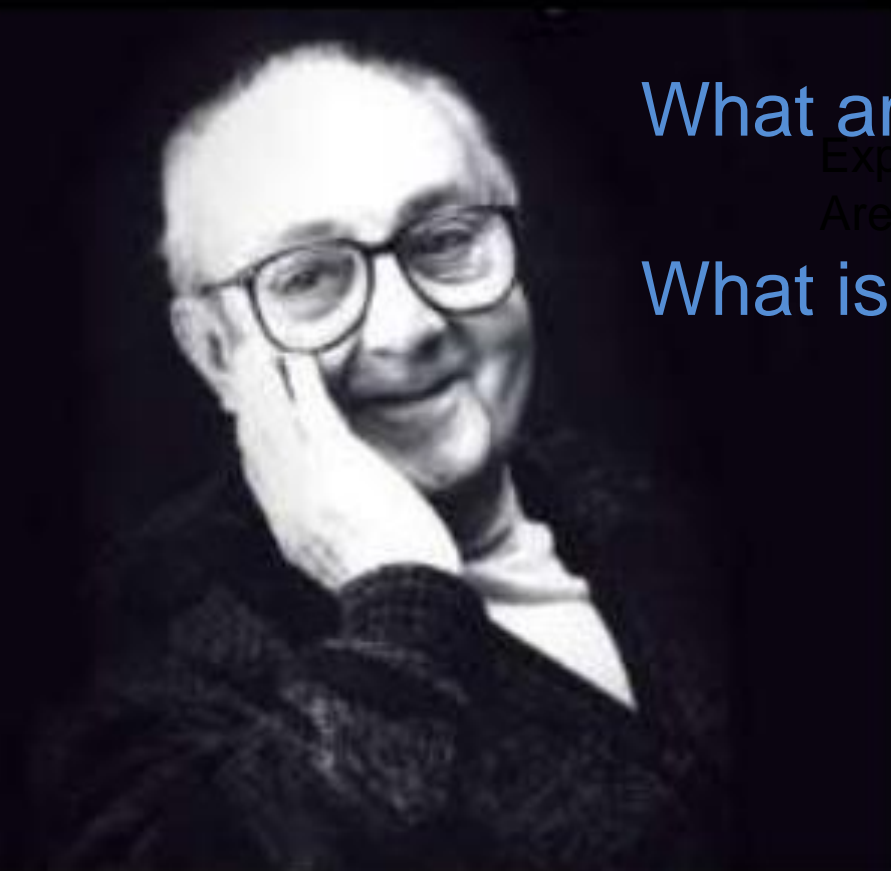


*All models are wrong  
but some are useful*



What are appropriate expectations?

Exp  
Are

What is the right measure of utility?

George E.P. Box

# Non-human Models of Neurodegenerative Disease



## Causes for Concern

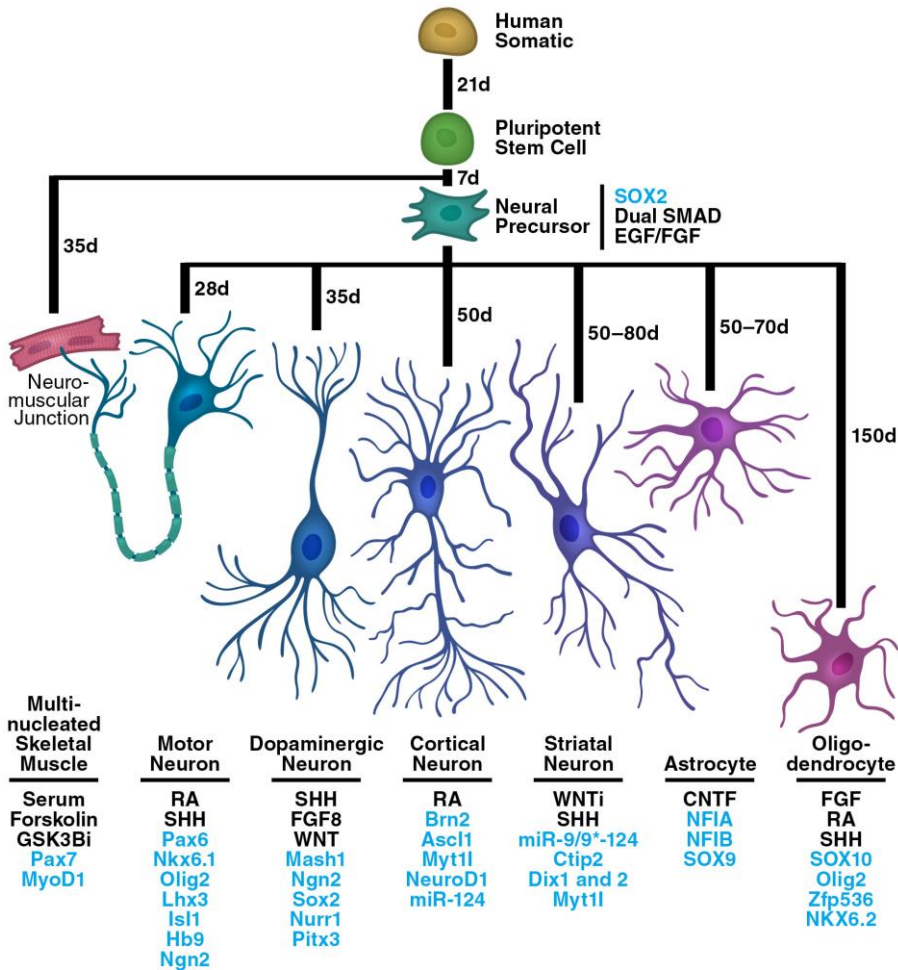
- Failure to reduce risk of drug development
- Species-specific differences in disease relevant genes, anatomy, physiology, aging, pharmacokinetics, etc... (Liao and Zhang (2008) *PNAS*)
- Will there ever be a “better” model?

## Unrealistic Expectations?

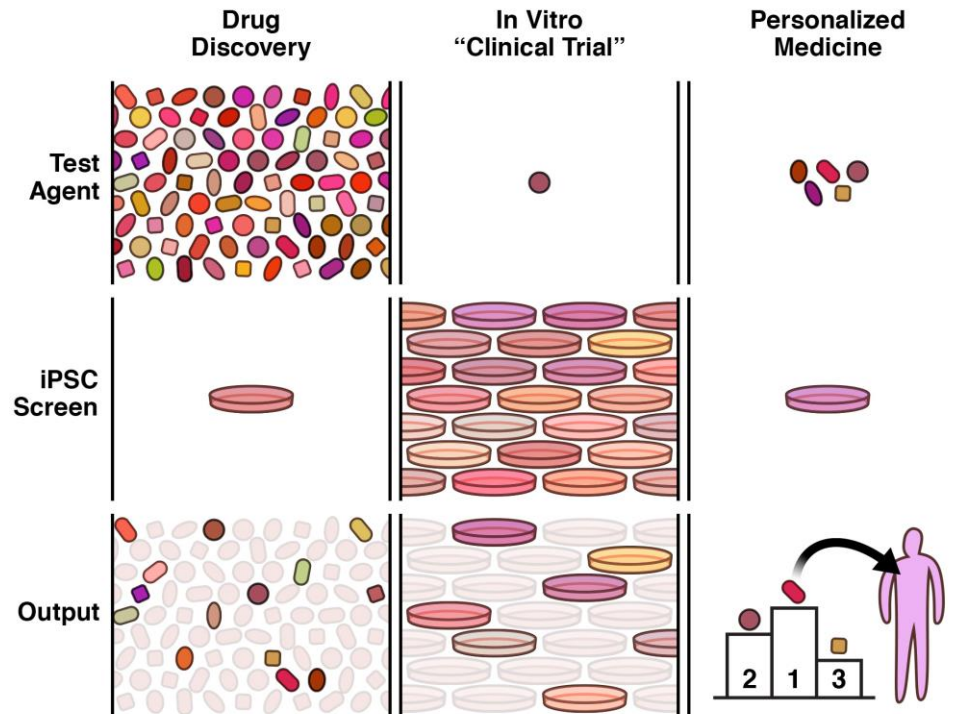
- Don't normally develop AD, PD, etc...
- Focus on recapitulation of human pathology may be misplaced, even counter productive
- Concept of complex adapted systems and “phenologs” (McGary et al. (2010) *PNAS*; Woods et al. (2013) *BMC Bioinformatics*)

*What is the power of a model to predict a clinical observation?*

# Can iPSC-based Disease Models Make Translation More Reliable and Help Deliver on the Promise of Precision Medicine?



- Human brain cells from patients with clinically defined disease
- iPSCs can be differentiated toward a variety of brain cell types
- Potential applications in drug discovery, patient stratification, etc...

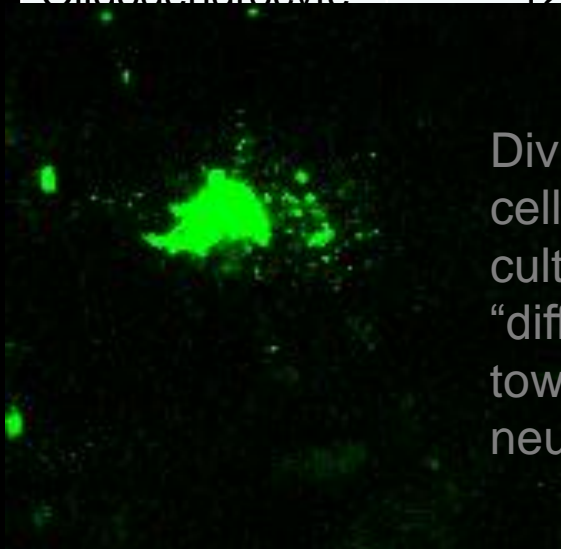


Haston and Finkbeiner, Ann. Rev. Pharm. Tox, (2016)

# Challenges: Heterogeneity by Accident or Design

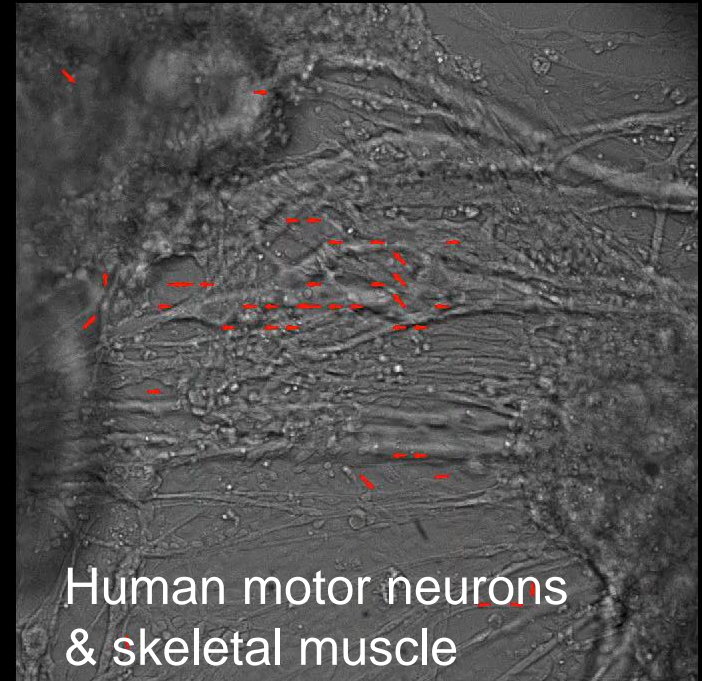
Cell type	Differentiation (weeks)	Efficiency (%)
Forebrain neuron	2	90-95
Dopaminergic neuron	4	40-50
Motor neuron	2-4	40-50
Striatal neuron	4	5-12
Astrocyte	5	60
Oligodendrocyte	12-13	10-20

- Differentiation protocols imperfect, long and complex
- Maturity, aging?
- HT Single cell analysis methods may be critical



Dividing Hb-9+  
cells in an iPSC  
culture  
“differentiated”  
toward motor  
neurons (day 18)

*Jeannette Osterloh and Kelly Haston*



Human motor neurons  
& skeletal muscle

# Automated Single Cell Analysis

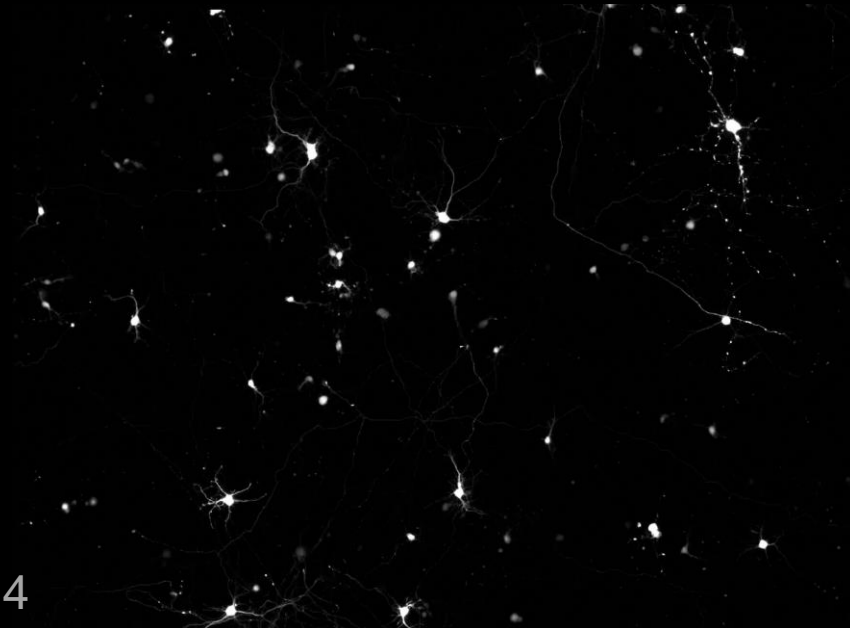


- 100-1000 fold more sensitive detecting phenotypes & drug effects

7 day movie



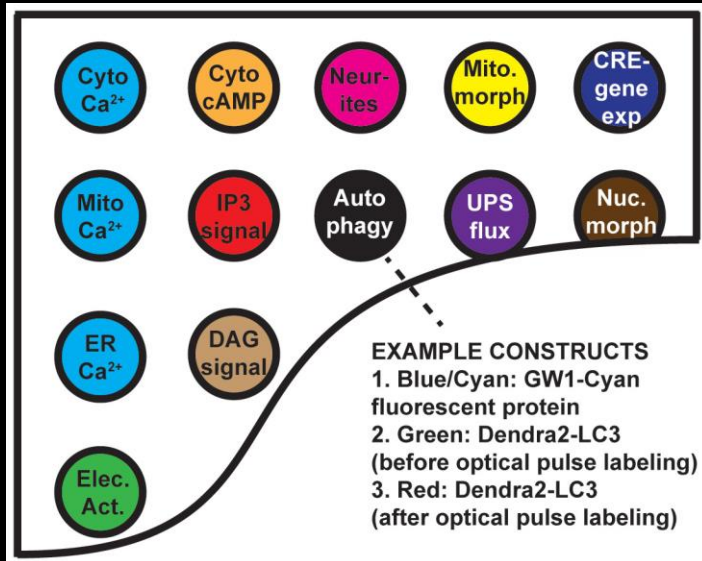
#thebrainbot





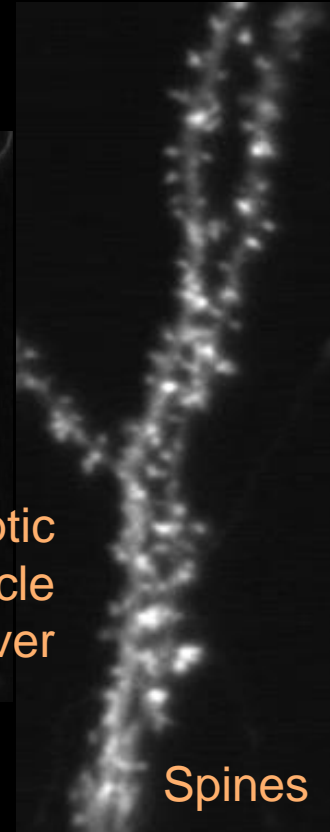
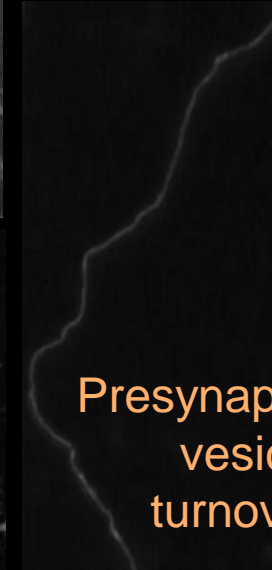
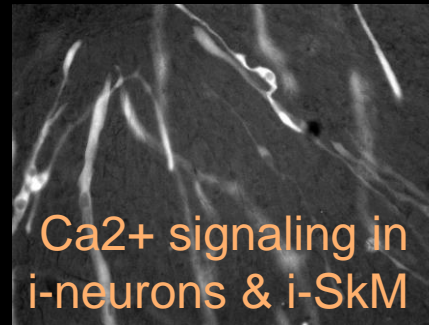
# High Throughput High Content Deep Phenotyping of iPSC Models

“Physical Exam” of the Cell



~ Array of 270 biosensors  
Multiplexed

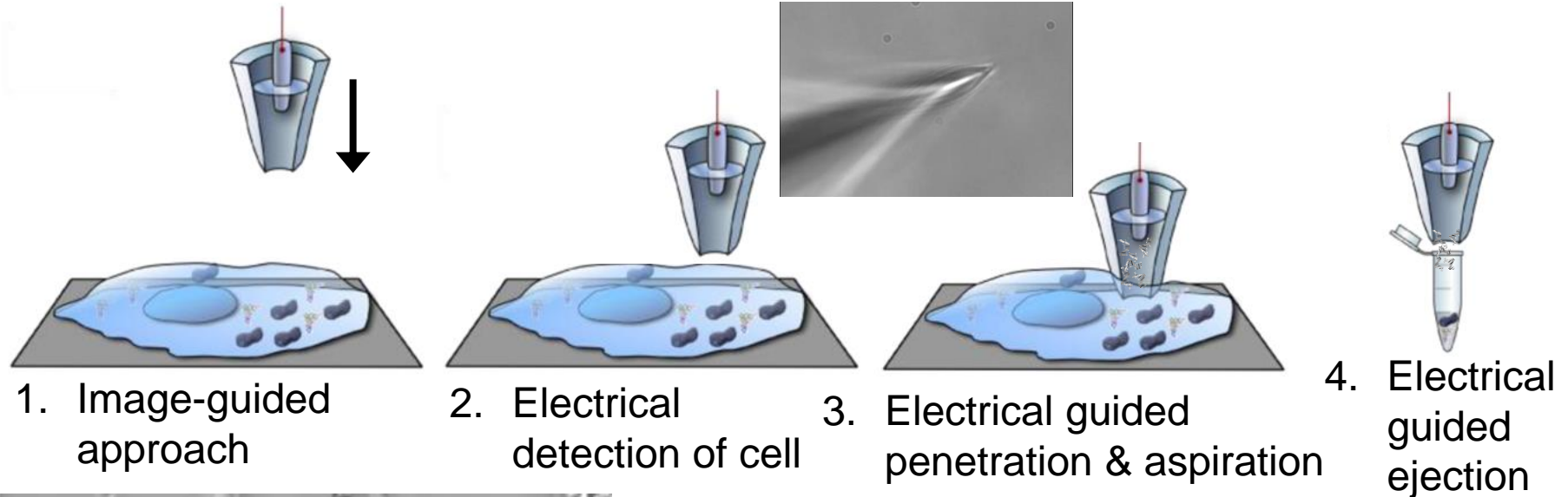
- “All optical” electrophysiology to stimulate and record
- Synapse structure / function
- Ca<sup>2+</sup> signaling
- Neurite extension and retraction



Neuron, 2015

- Mitochondrial structure, trafficking
- Bioenergetics, metabolism
- Proteostasis flux, autophagy
- DNA damage and repair
- Protein aggregation, metabolism

# Fully Automated Dynamic Single Cell Transcriptomics



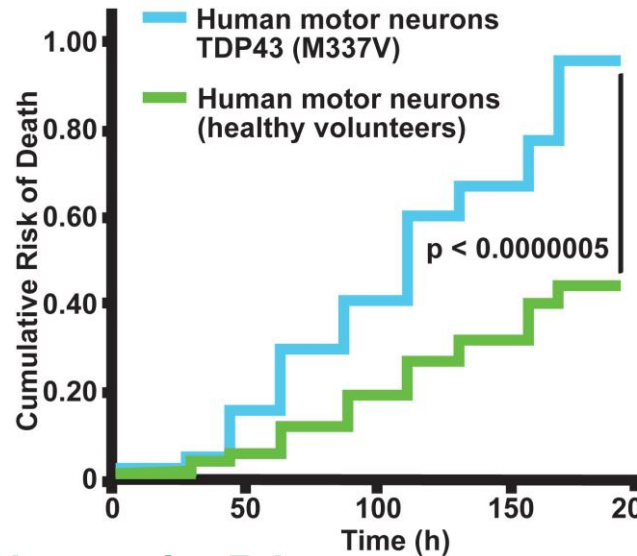
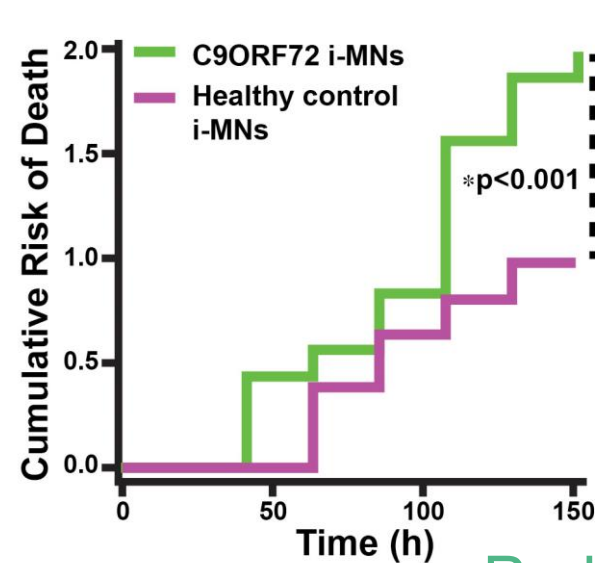
- Relate to patient transcriptomics
- Understand molecular basis for phenotypic differences according to
  - Cell type
  - Adaptive response
  - Perturbagen
- Target ID from genetic screens

Differentiated human  
striatal i-neurons

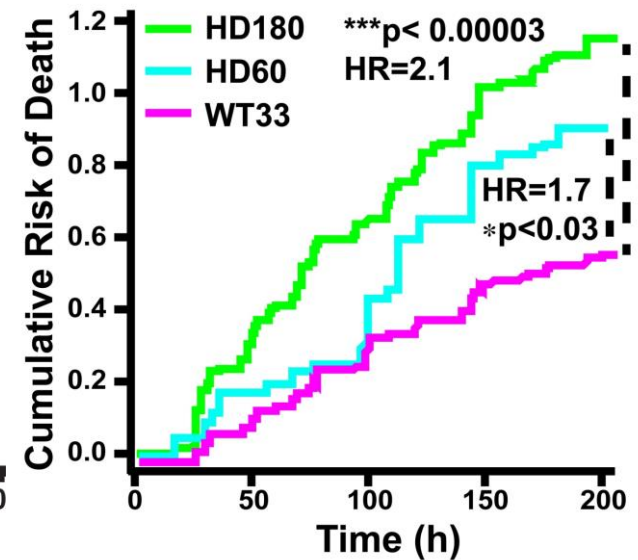
Seeger et al. (2012) *Nanoscale*; Actis et al. (2014) *ACS Nano*

# Human Disease Model Phenotypes – No Stressors

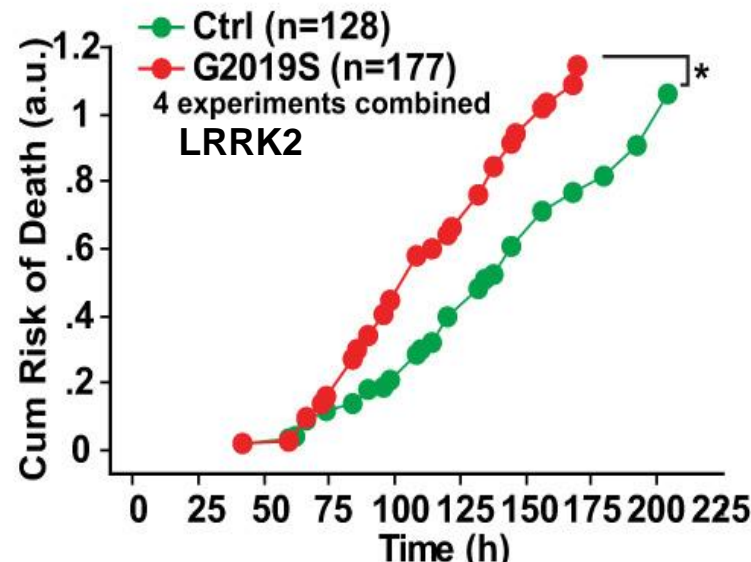
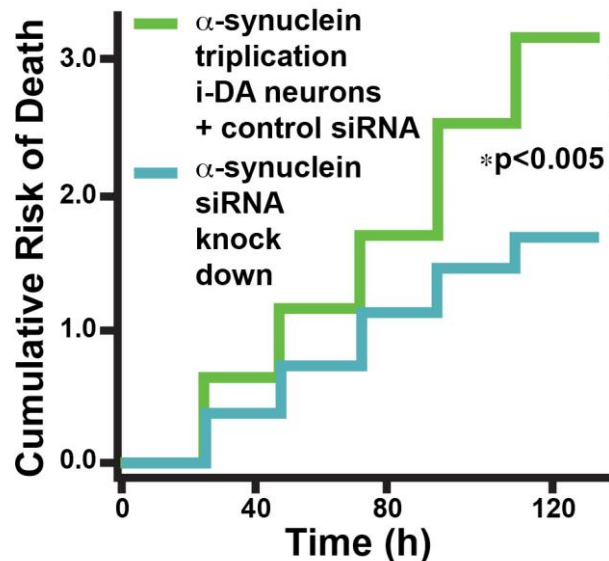
## ALS / Frontotemporal dementia



## Huntington's Disease



## Parkinson's Disease

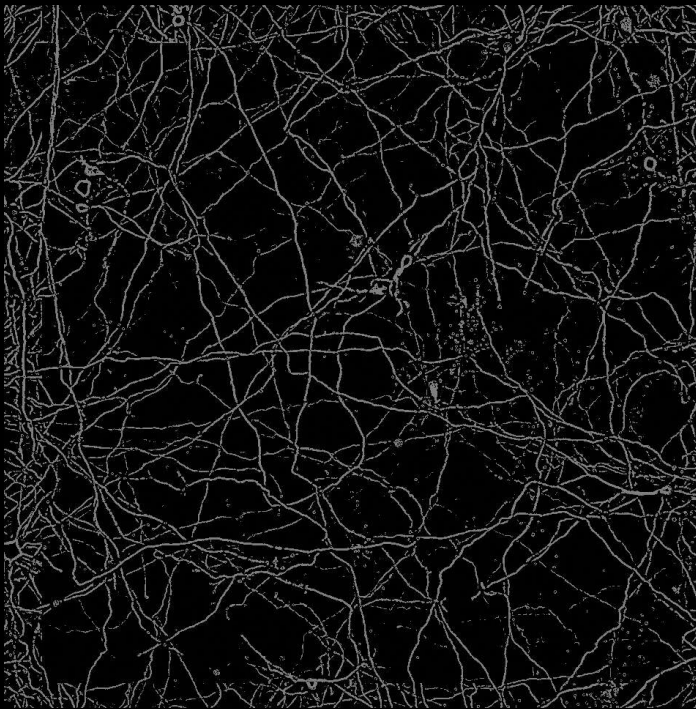


## Disease-relevant phenotypes:

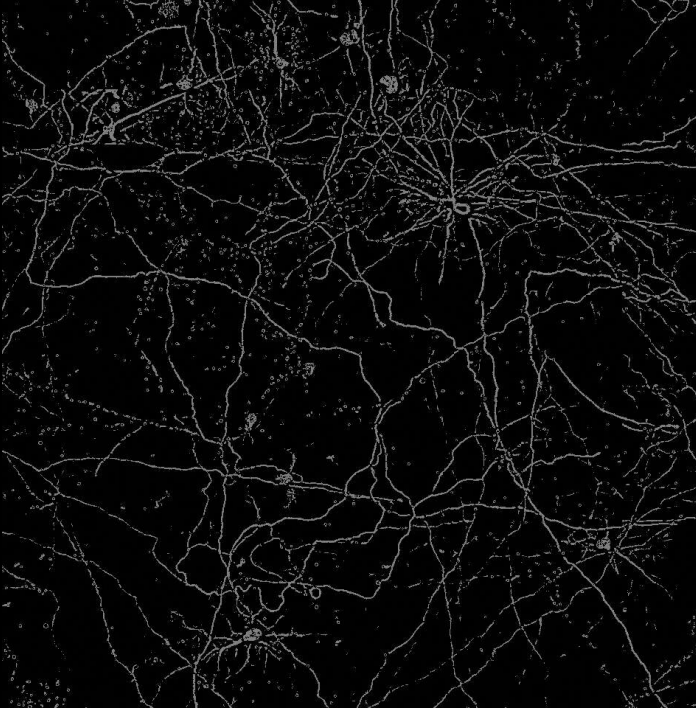
- Survival
- Neurite length
- Calcium signaling
- Glial phenotypes
- Electrophysiology changes
- Gene profiling changes



Healthy Control



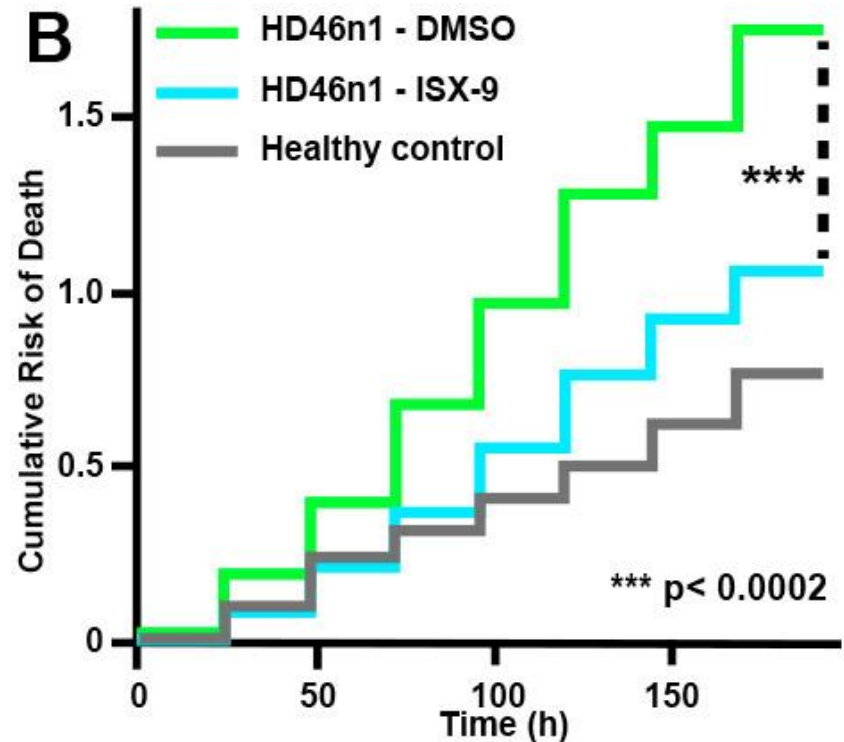
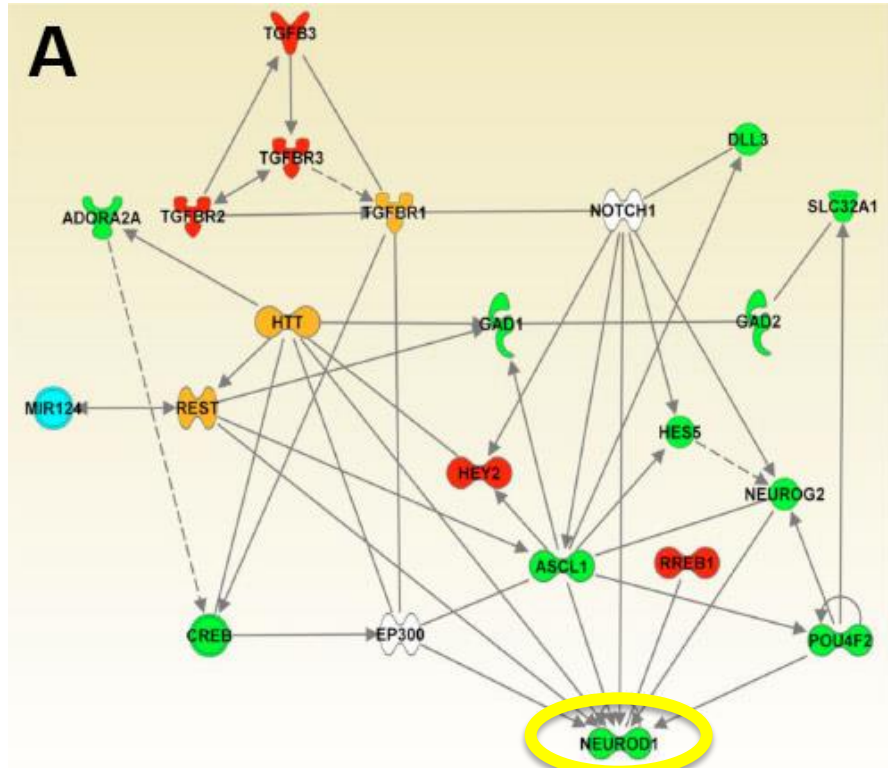
TDP43-ALS



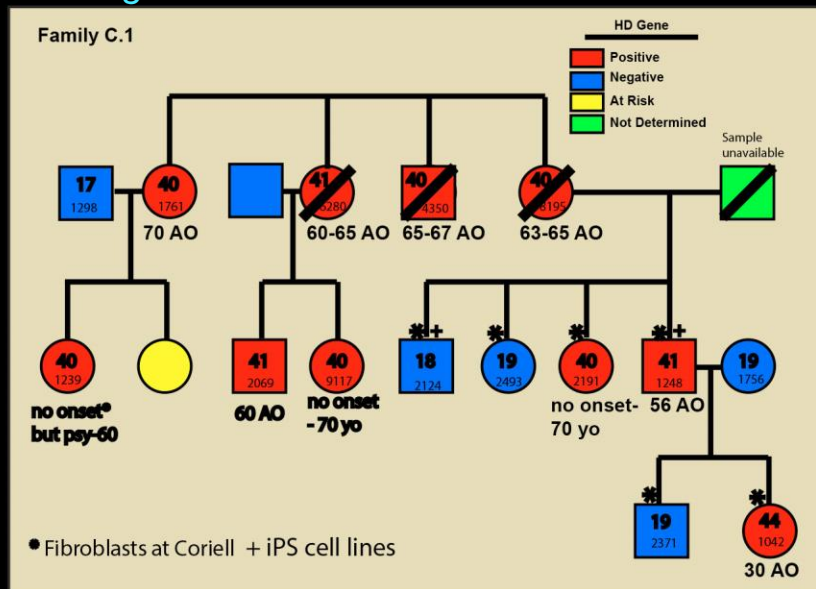
## Applications of Deep Learning to See Things Better:

Automated Neurite Analysis &  
“Diagnose a Well”

# Roles of iPSCs in HD – Mechanisms of Pathogenesis



- Whole transcriptome analysis suggests a major deficit in several developmental pathways
- NeuroD and its downstream targets significantly downregulated in HD i-neurons
- Survival and physiology deficits mitigated by a NeuroD activator, ISX-9



## Whole Genome Analysis of HD Families:

**A Rare Genetic Variant of a Proteastasis Gene is Associated with Delayed Symptom Onset**

**Julia Kaye, unpublished**  
Alicia Holloway, Stacia Wyman

# Answer ALS: Largest most comprehensive clinical and biological assessment of ALS



- **Part 1.** A comprehensive and longitudinal deep clinical data set from at least 1,000 ALS patients
  - ~2 year followup (Guid assignment); 5 visits
  - Extensive deep clinical data: EMG, clinical, MRI, pulmonary
  - Biofluids and tissue: Blood, CSF (~10%), Autopsy (~10%), DNA, PBMC
  - Personal monitoring device: 24/7 clinical data
- **Part 2.** Generate iPSC cells from every patient (after visit 1)
  - iPSC Motor neurons and astroglia
- **Part 3.** Perform comprehensive biological analytics these human brain cells:
  - Whole genome, Transcriptome, proteome, epigenome, metabolome, lipome), robotic serial neuronal analytics.
- **Part 4.** Data Analytics:
  - clinical/omics based sub
  - Pathways
  - Biomarkers
  - *Future clinical trials based on data sets/biologicc; New biomarkers*
  - *New ALS Druggable Pathways*





# Private sector environment for proceeding in absence of animal models to predict efficacy

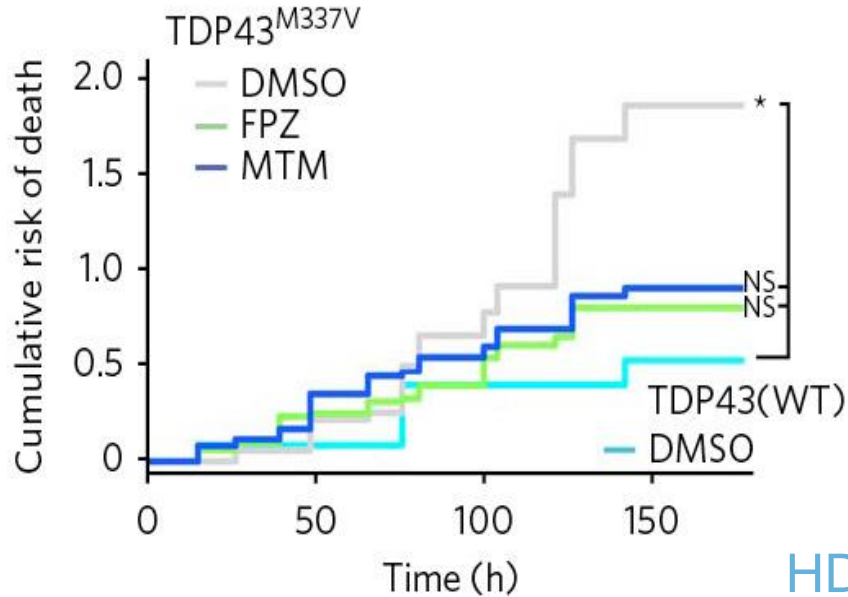


~ 50% tell us they are willing to proceed without in vivo efficacy if they have:

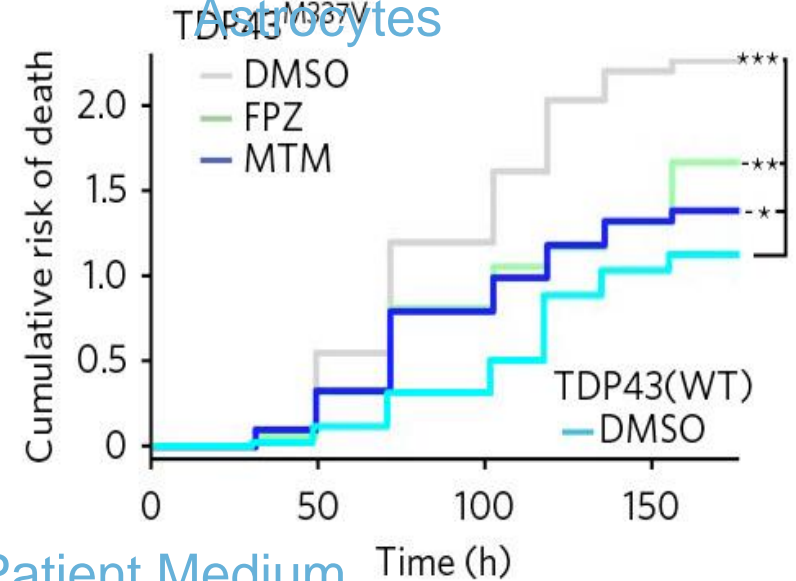
1. Efficacy in iPSCs
2. *In vivo* tox
3. Target engagement biomarker in humans

# Autophagic Induction Protects Patient Human Neurons from Degeneration

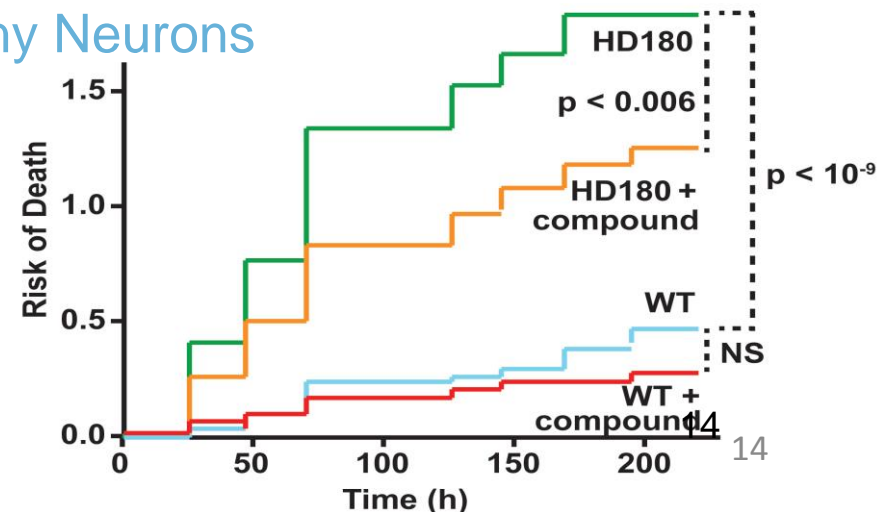
## ALS Patient Motor Neurons



## ALS Patient Astrocytes



## HD Patient Medium Spiny Neurons



- Accelerate the neuronal clearance of disease proteins in i-neurons
- Mitigate cytopathology
- Lead optimization
- Nanomolar potency, BBB penetrant

Yush Goyal, Ashkan Javaherian, Julia Kaye

Tsvetkov et al. Proc. Natl. Acad., 2010; Tsvetkov et al. Nat. Chem. Biol., 2013; Barmada et al. Nat. Chem. Biol., 2014