

# HyperFusion: Imaging-Tabular Data Integration for Predictive Modeling in Healthcare

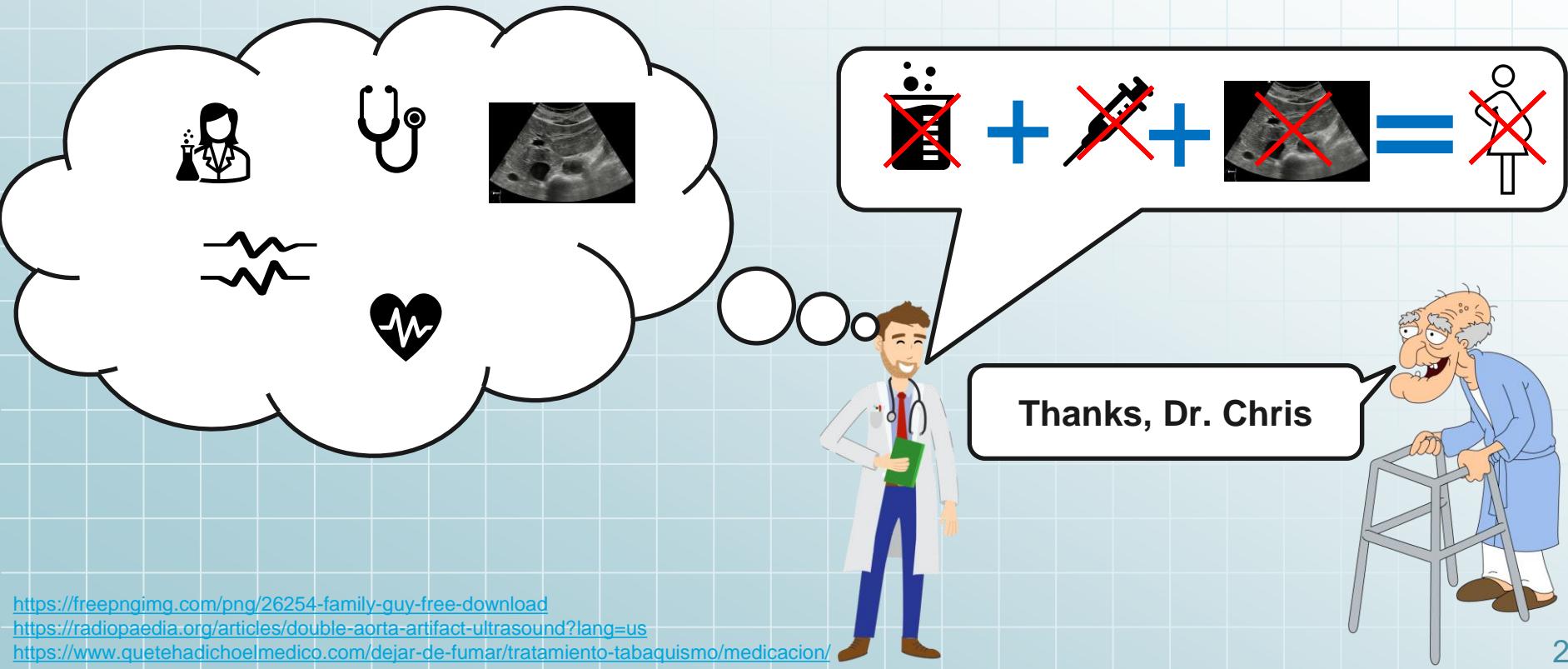
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# Multi-Modal Fusion

- Doctors intuitively performs multimodal Data fusion

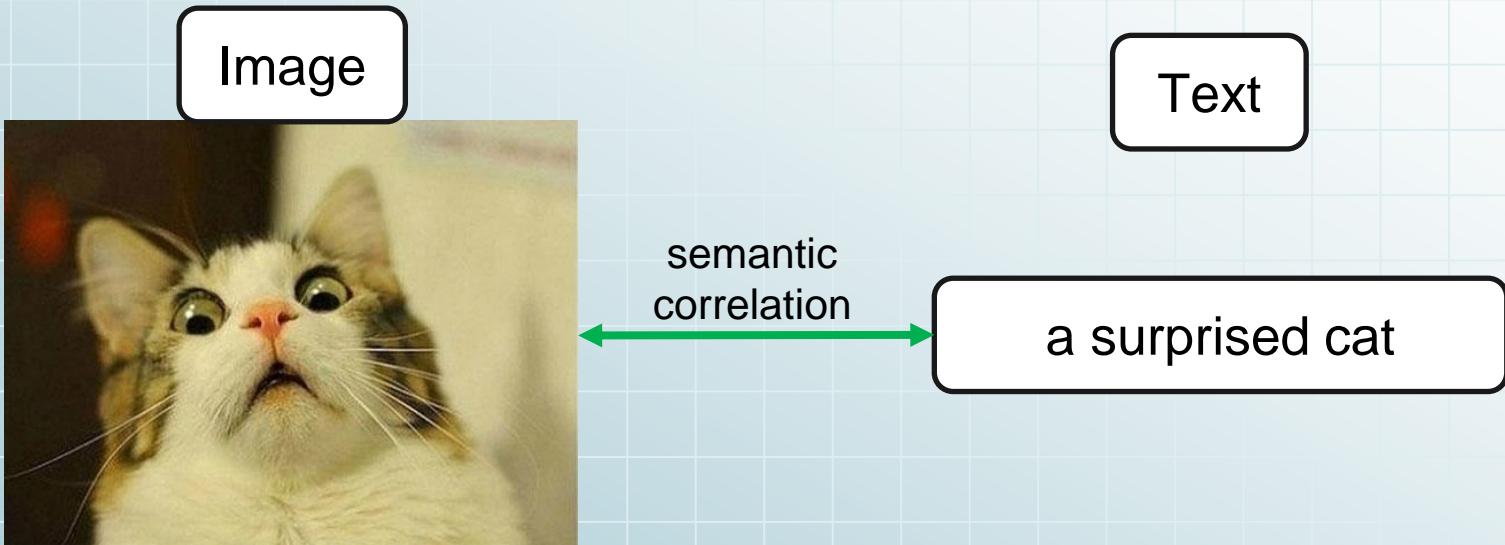


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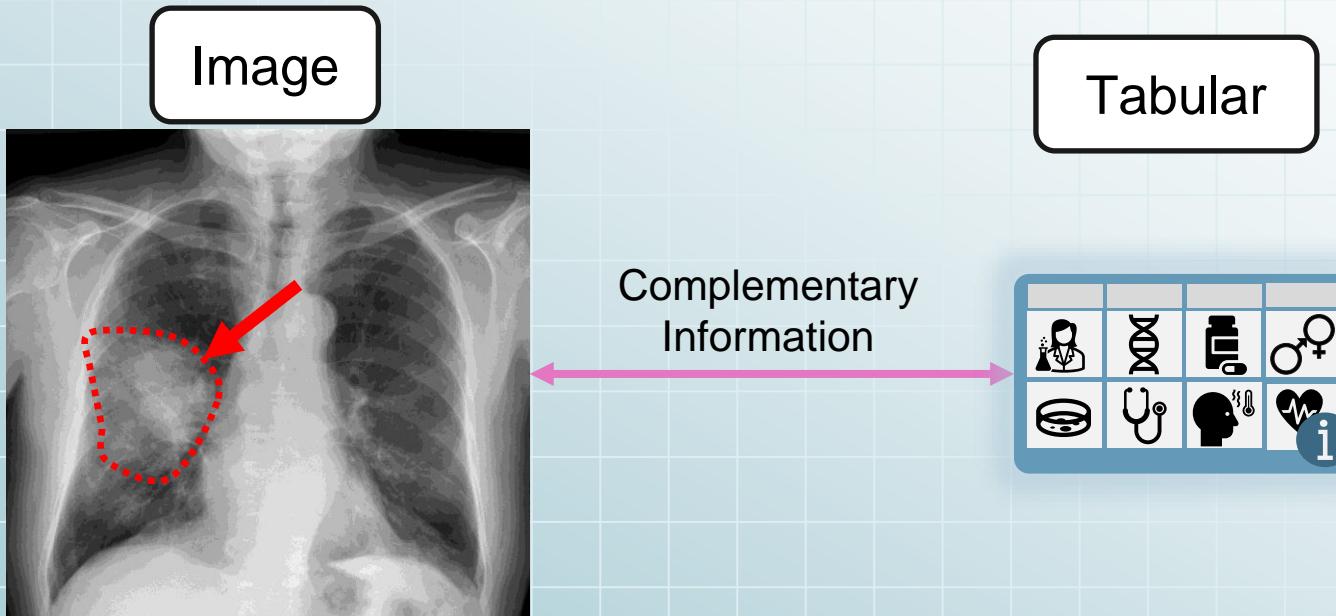
<https://radiopaedia.org/articles/double-aorta-artifact-ultrasound?lang=us>

<https://www.quetehadichoelmedico.com/dejar-de-fumar/tratamiento-tabaquismo/medicacion/>

# Text-Imaging Data Fusion

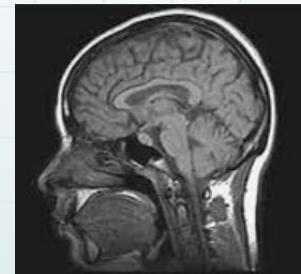
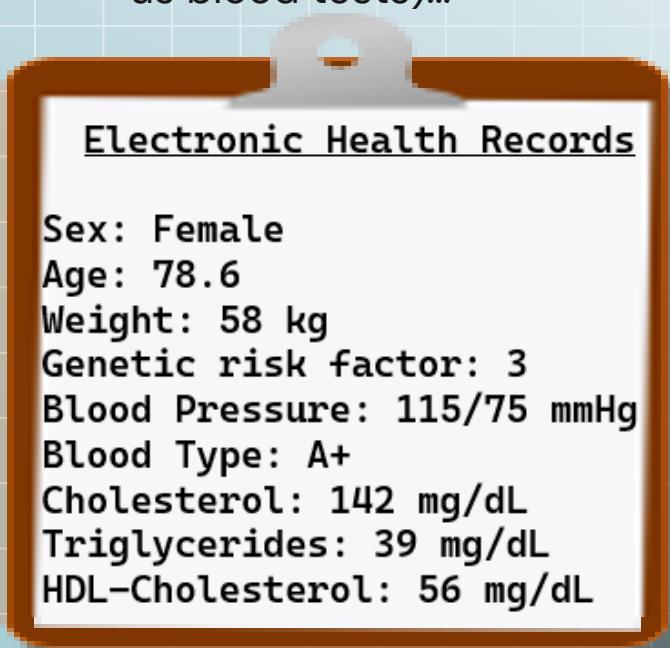


# Tabular-Imaging Data Fusion



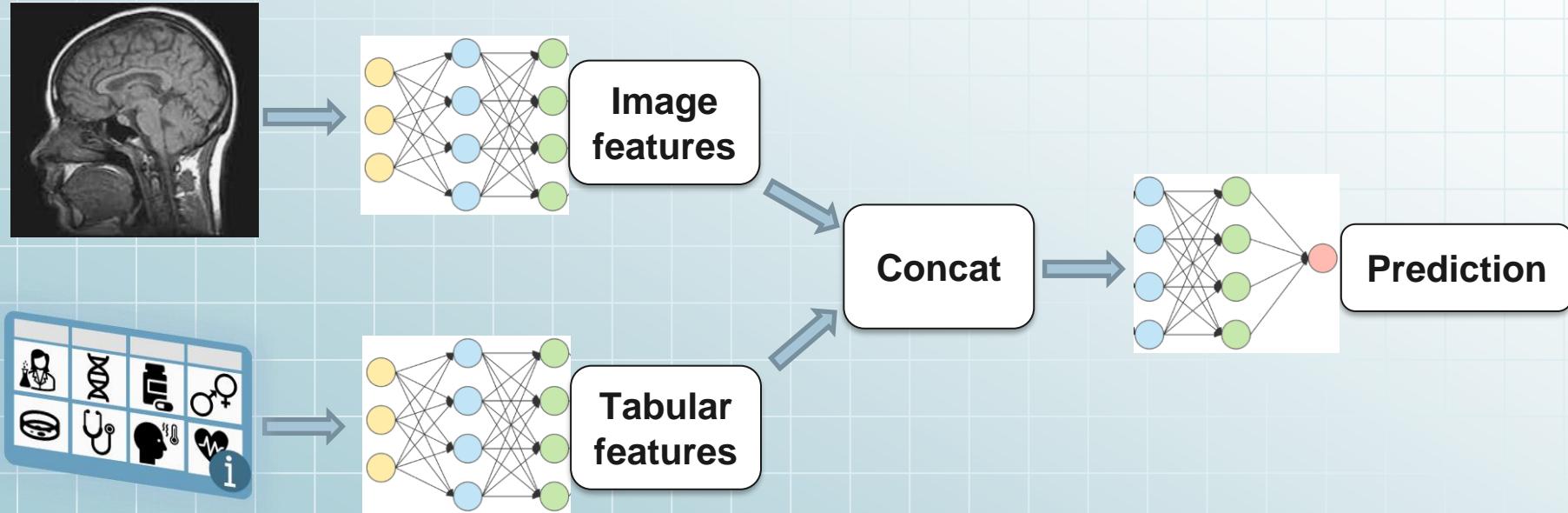
# Tabular-Imaging Data Fusion

- Tabular data derived from Electronic Health Record (EHR)
- Contains clinical info, genetics, demographic info, and lab results (such as blood tests)...



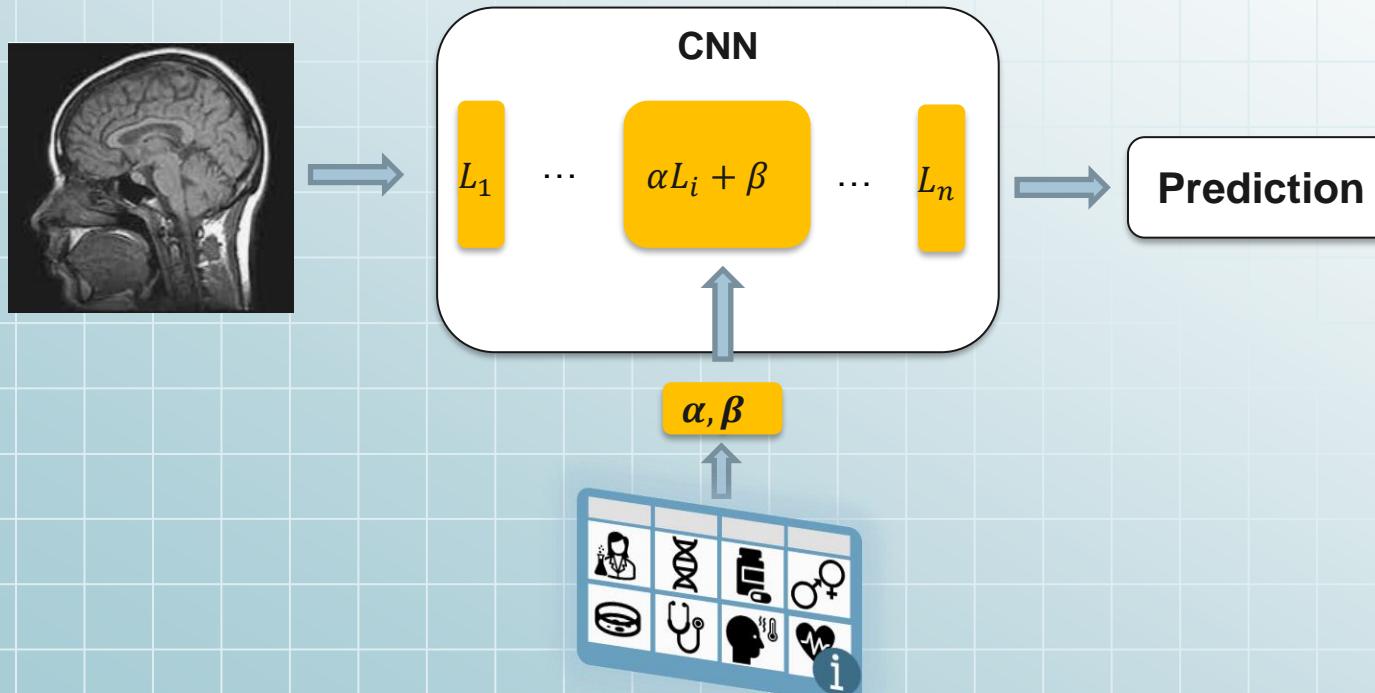
# Images - Tabular Data Fusion Approaches

- Straightforward approach – concatenation:

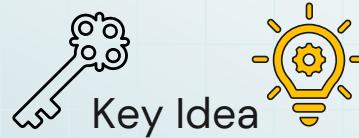


# Images - Tabular Data Fusion Approaches

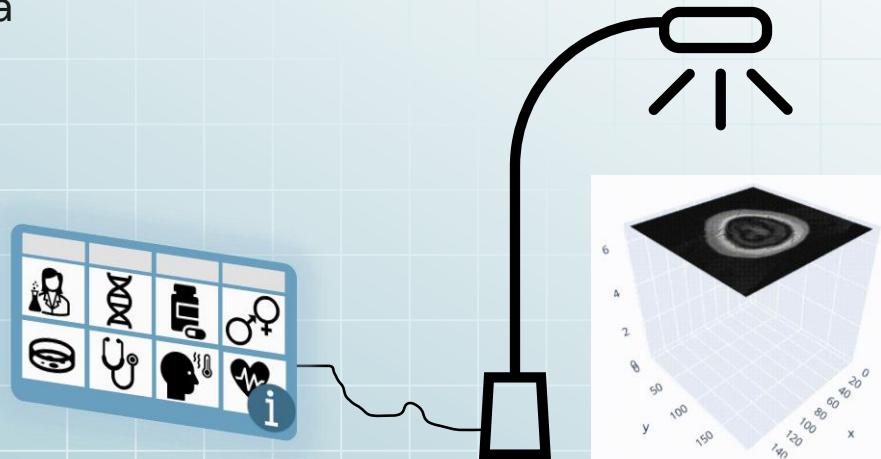
- Affine transformation of the imaging features



# Use Hypernetworks For The Tabular-Imaging Fusion

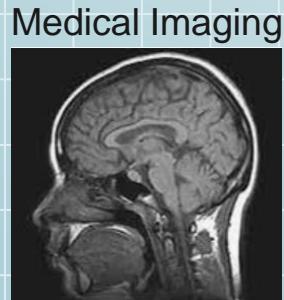


Use Hypernetworks to “look” at the image in the light of the tabular data



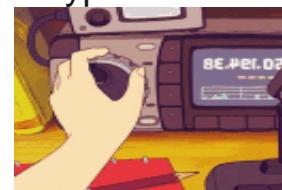
# Hypernetworks

Clinical Tabular Data

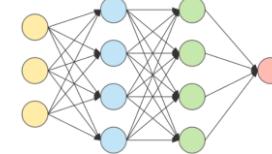


Network compound

Hypernetwork



Primary Network



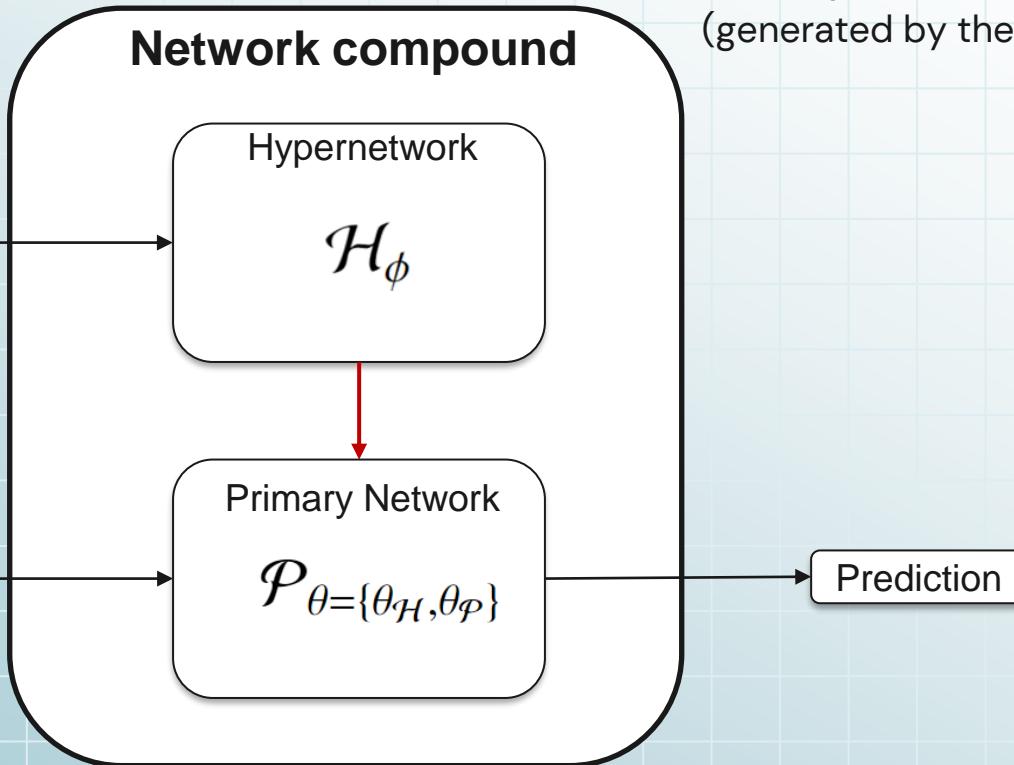
Prediction

# Hypernetworks

Clinical Tabular Data



Medical Imaging



Parameters:

- $\phi$  - Hypernetwork's internal
- $\theta_P$  - Primary network's internal
- $\theta_H$  - Primary network's external  
(generated by the Hypernetwork)

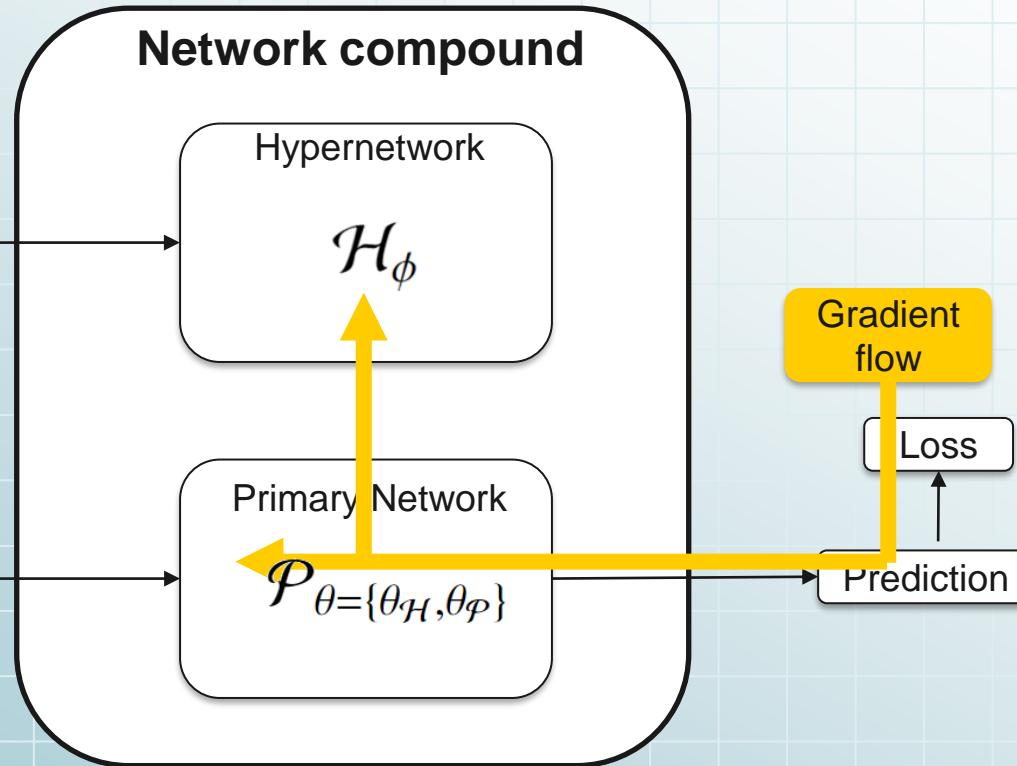
# Hypernetworks

Backpropagation – gradient flow

Clinical Tabular Data

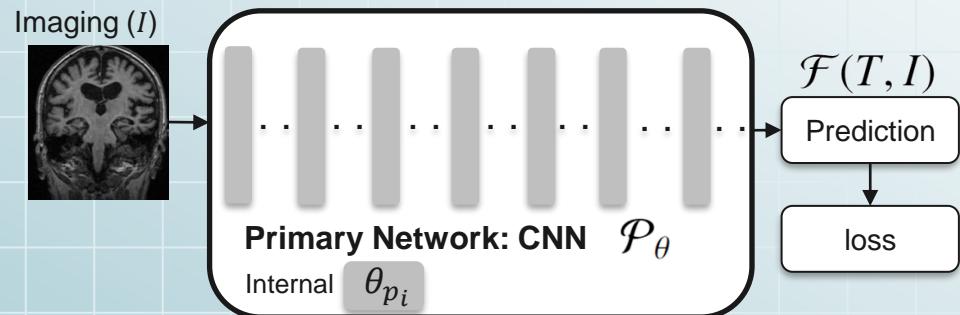


Medical Imaging



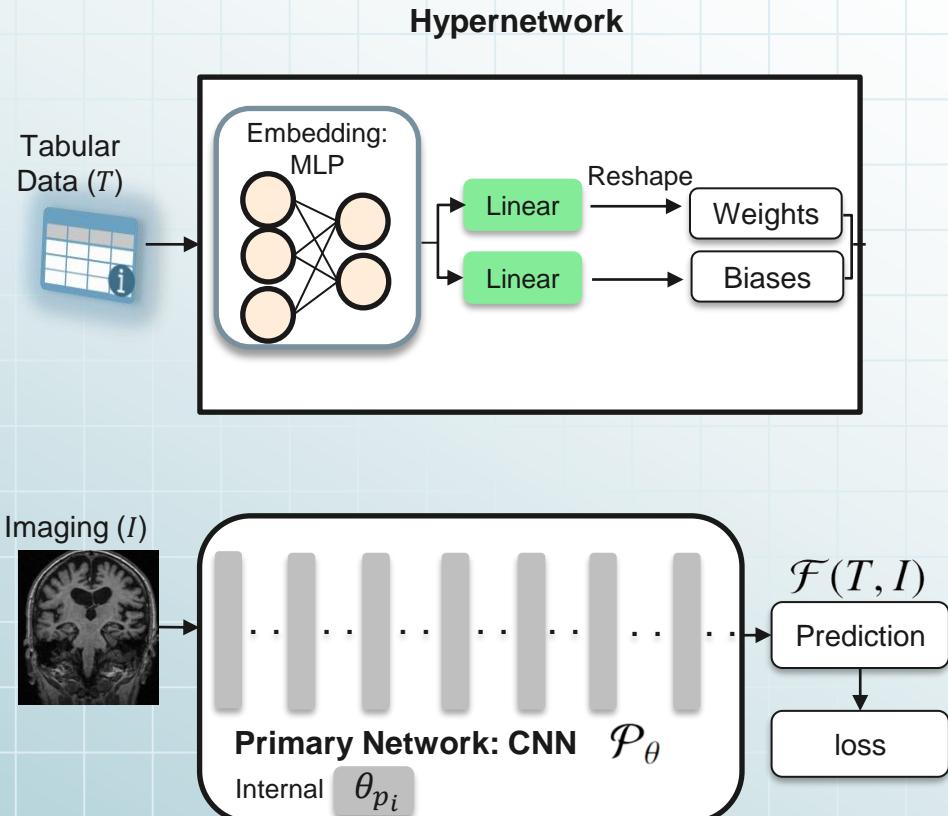
# HyperFusion General Architecture

- Primary network – any network  
(we chose CNN)



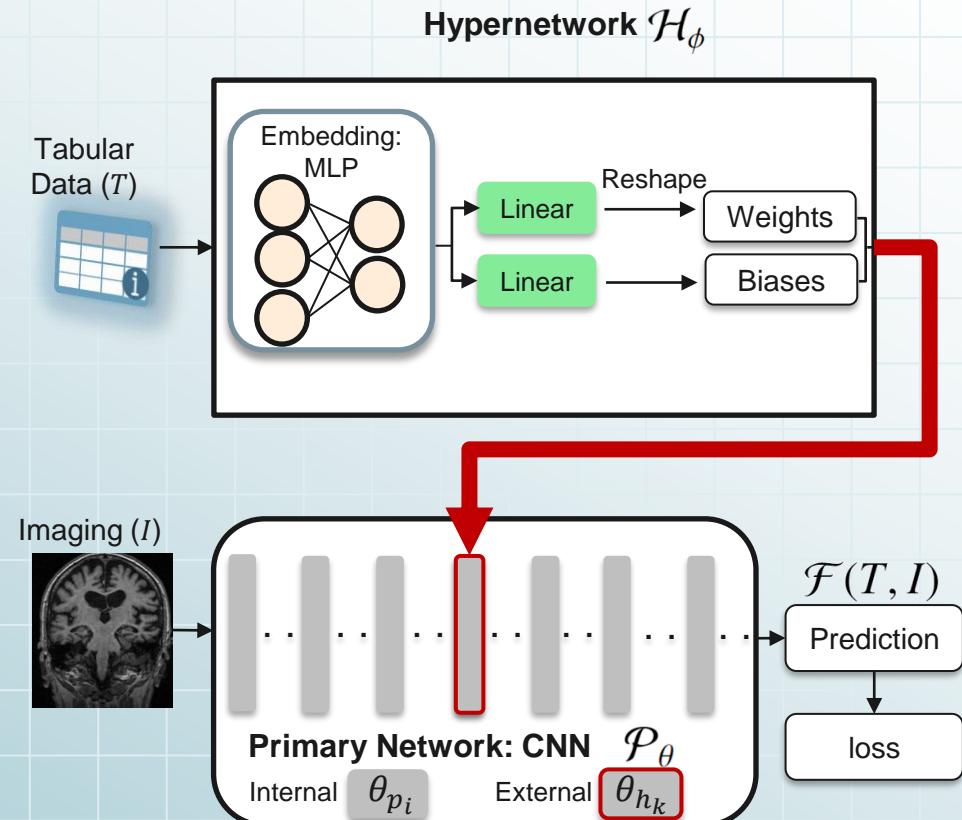
# HyperFusion General Architecture

- Primary network – any network (we chose CNN)
- Hypernetwork – any network (we chose MLP)



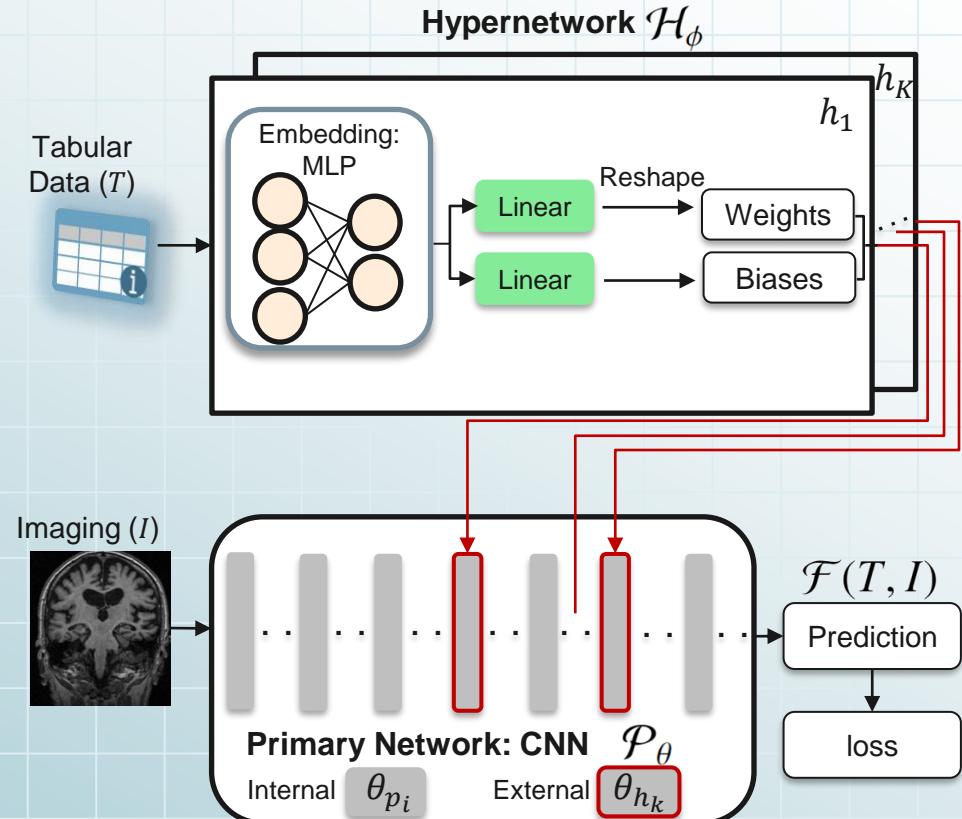
# HyperFusion General Architecture

- Hypernetwork – any network (we chose MLP)
- Weights and Biases for a designated layer (any type of operation)



# HyperFusion General Architecture

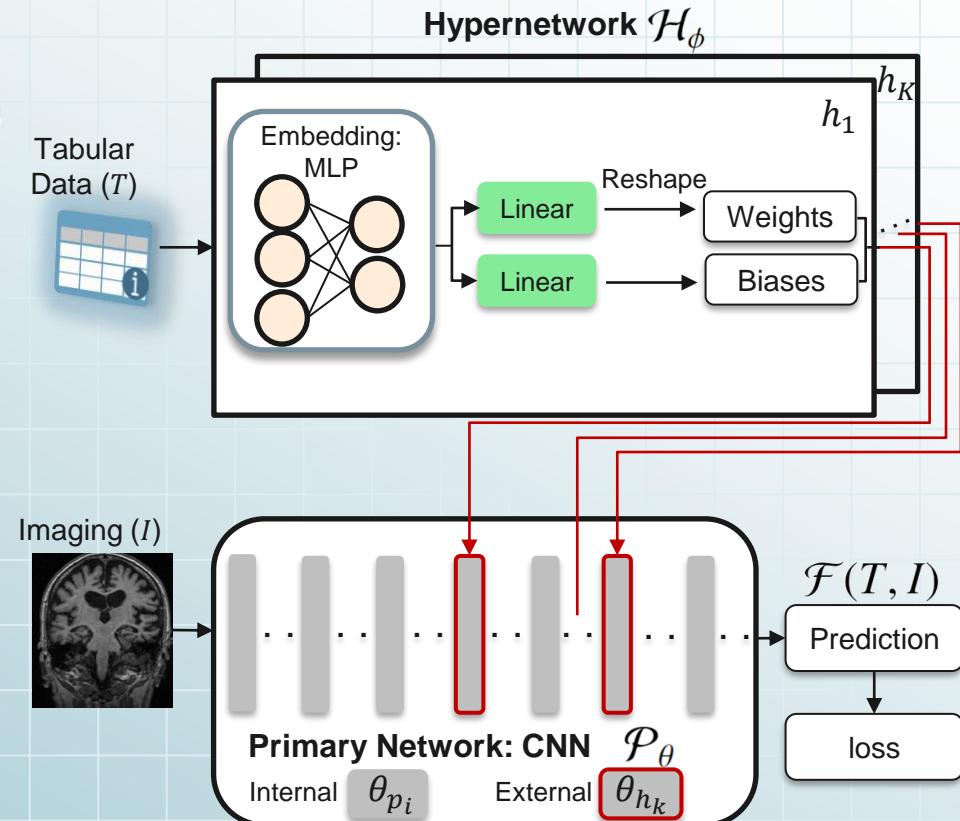
- K hypernetworks for K designated layers
- Which layers should be connected to the hypernetwork?
  - Usually high-level features layer



# HyperFusion General Architecture

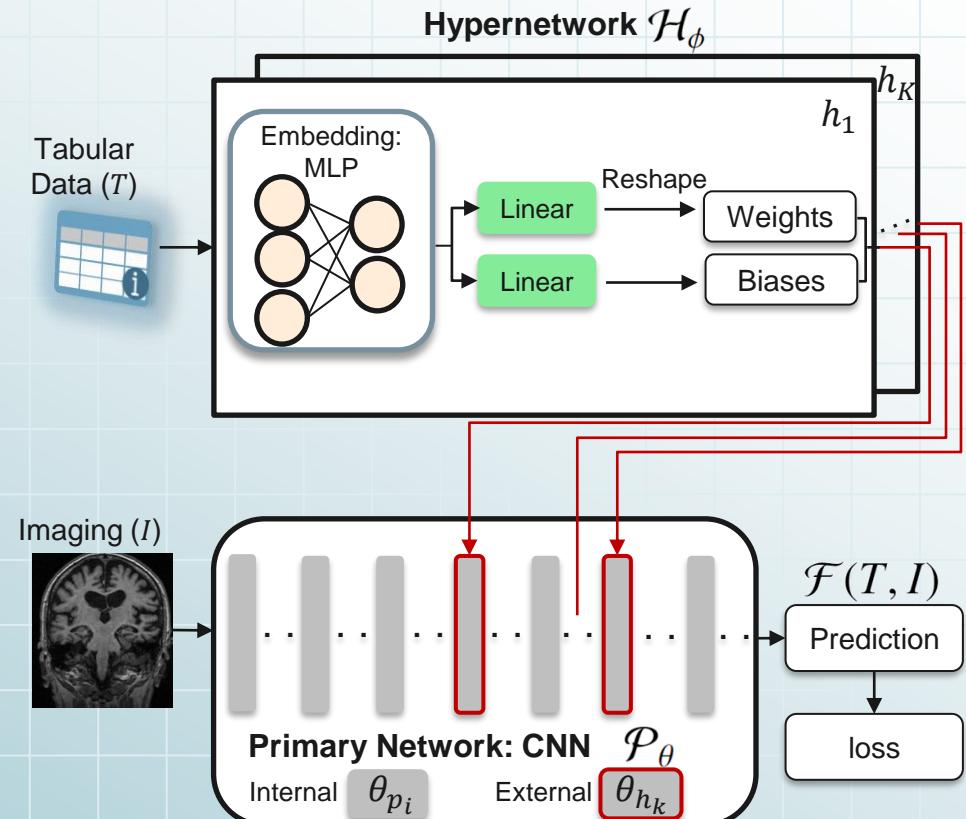
- Loss

$$\mathcal{L}(y, \mathcal{F}(T, I)) = \mathcal{L}_{\text{task}}(y, \mathcal{F}(T, I)) + \mathcal{L}_{\text{reg}}(\{\phi, \theta_{\mathcal{P}}\})$$



# Challenges and Other Important topics

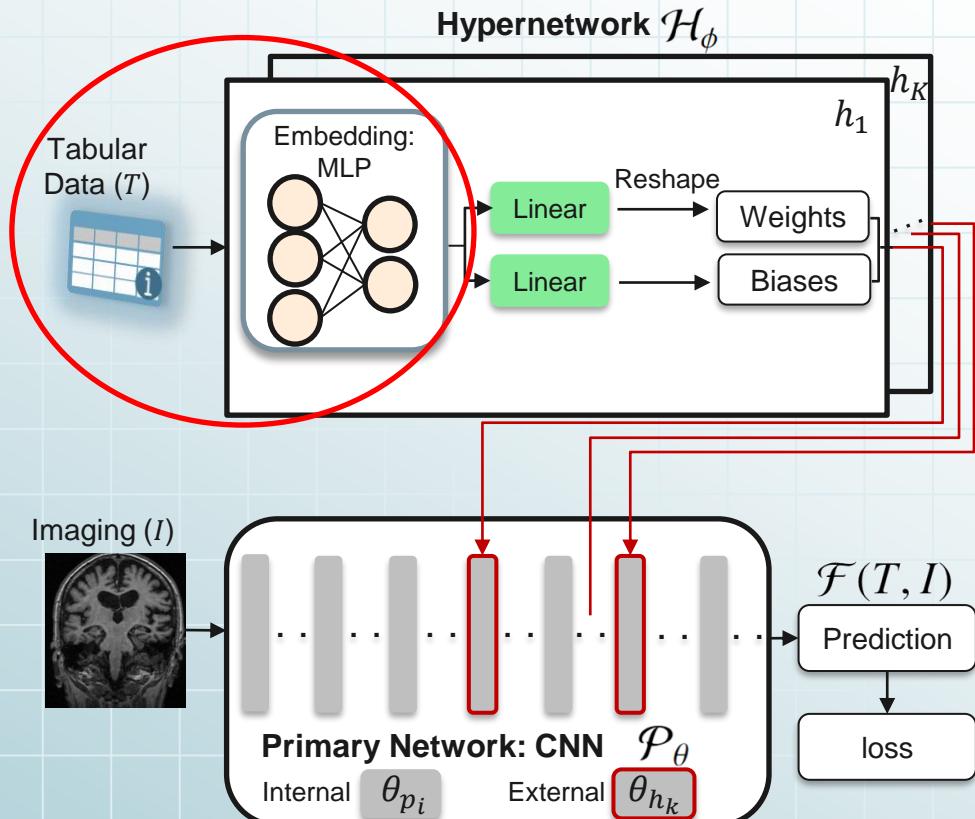
- Embedding
- Weights initialization
- Missing values



\*More details in the paper

# Challenges and Other Important topics

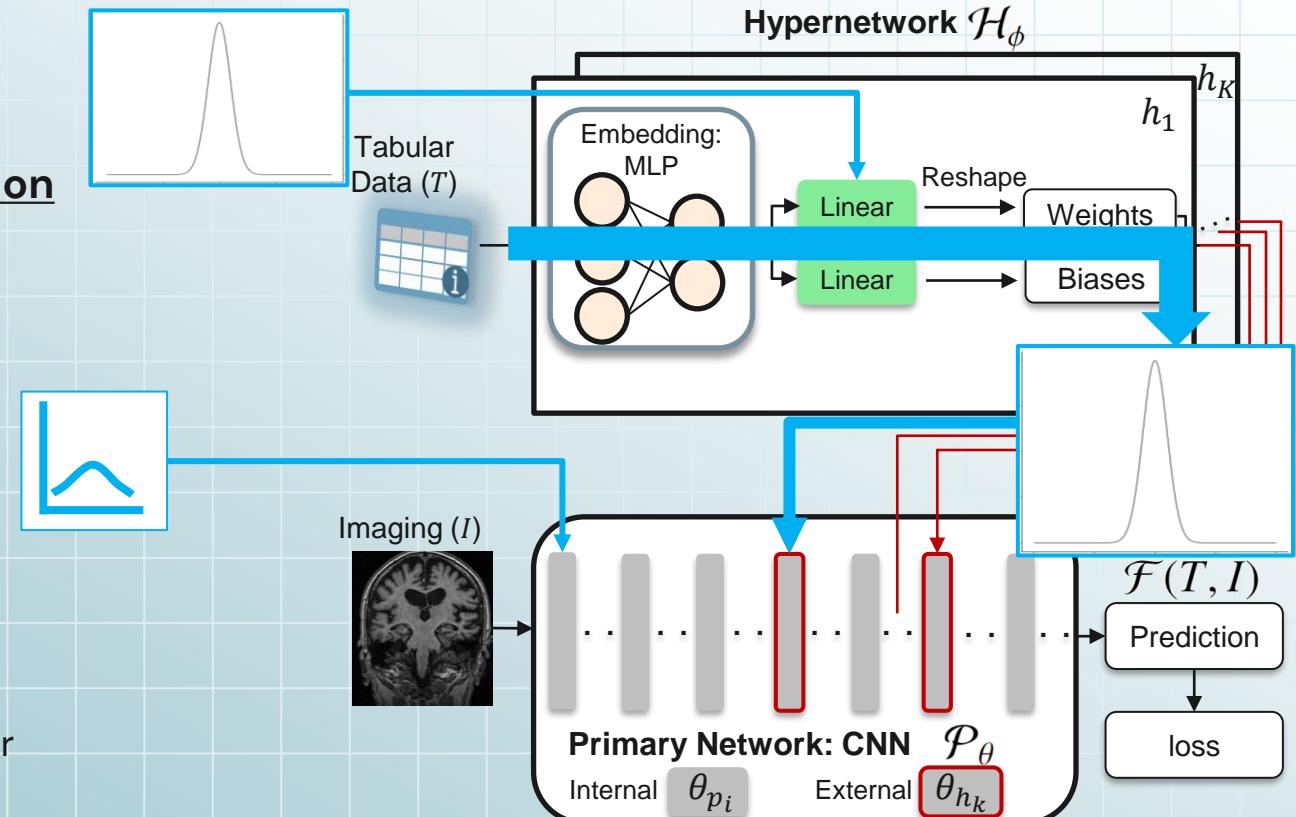
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# Challenges and Other Important topics

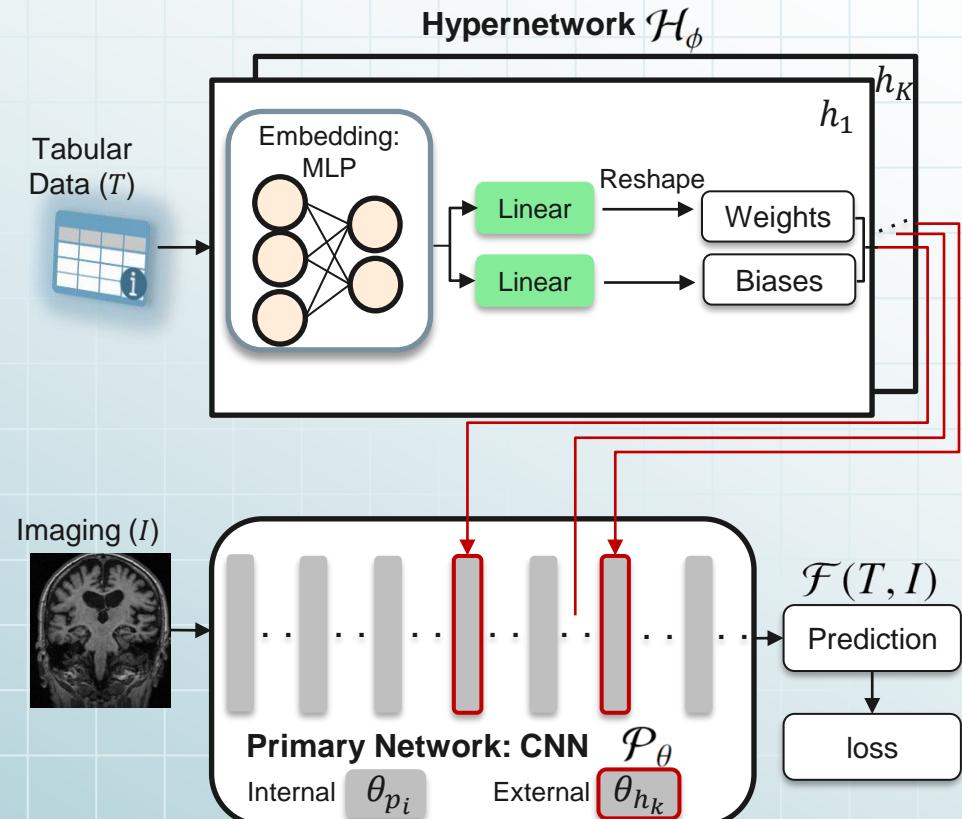
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# Challenges and Other Important Stuff

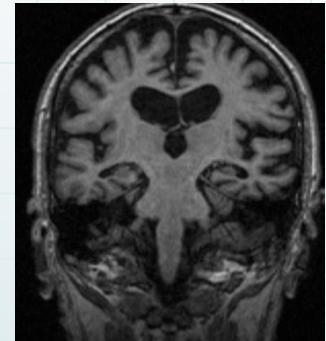
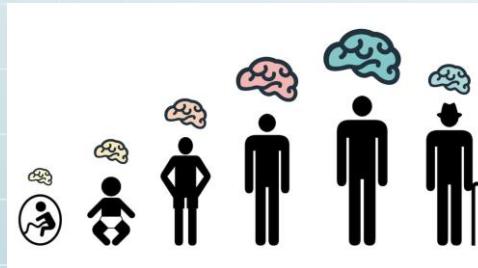
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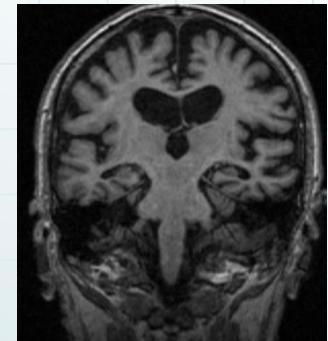
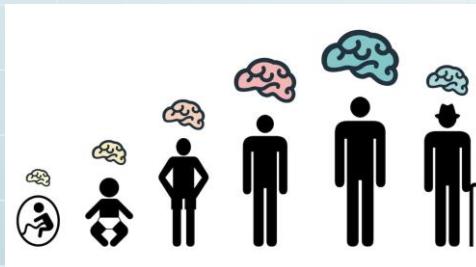
# Aging and Alzheimer's

1. Brain Age prediction (regression) – MRI and single tabular attribute (sex)

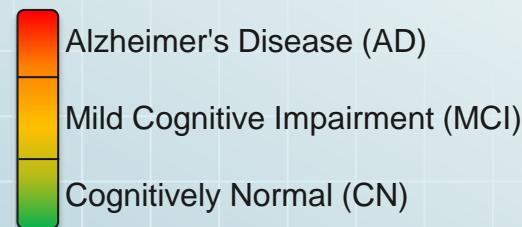
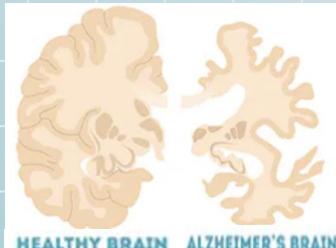
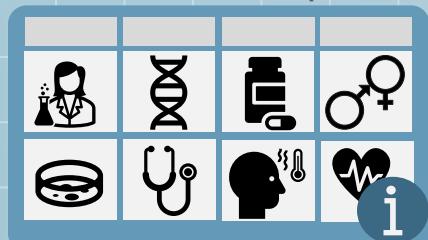


# Aging and Alzheimer's

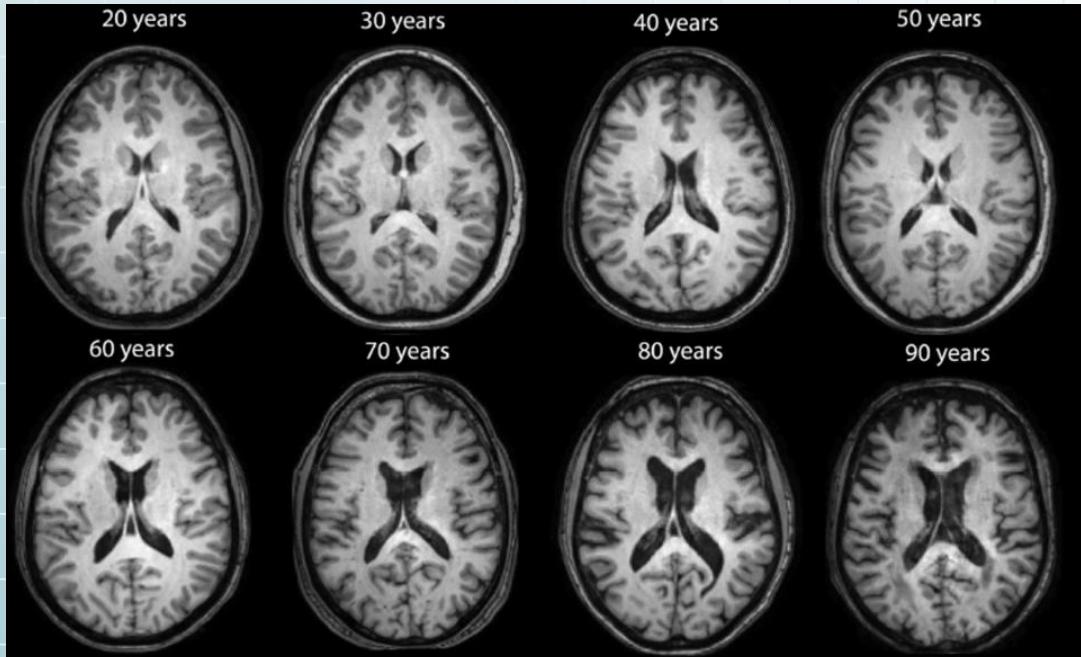
1. Brain Age prediction (regression) – MRI and single tabular attribute (sex)



2. Alzheimer Disease Diagnosis (classification) – MRI and multiple tabular parameters (different attribute types, missing values...)

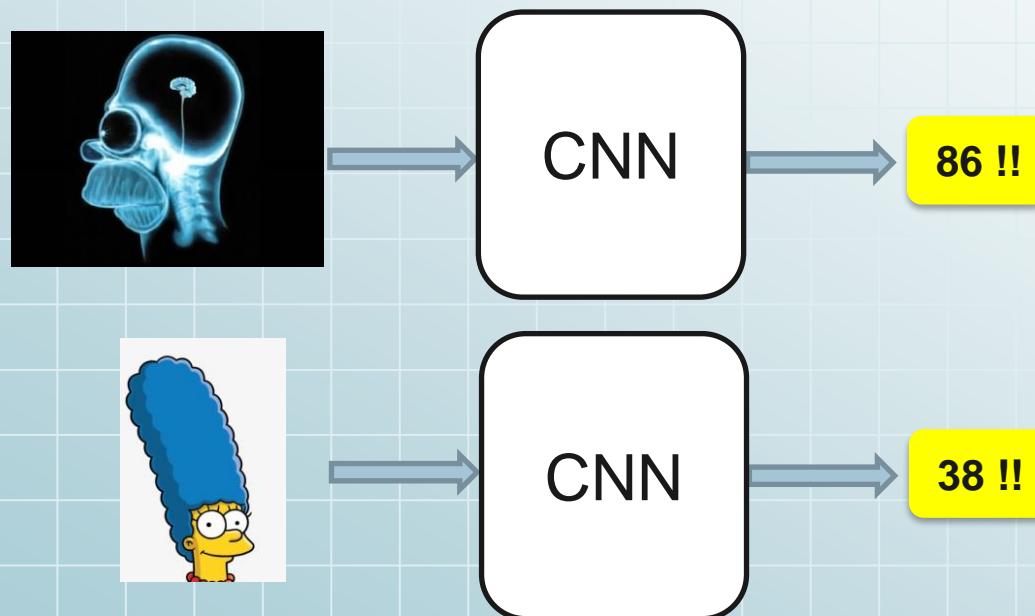


# Brain Age Prediction



# Brain Age Prediction

- Do males and females age differently?

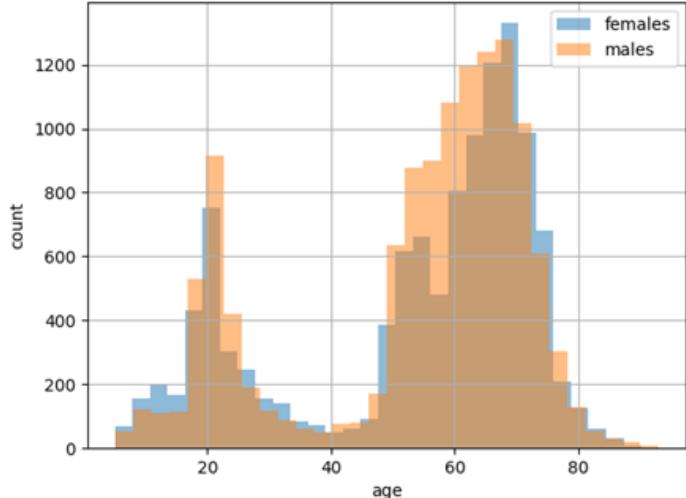


# The Data

- ~27K brain images from 19 different datasets
- 49% Males, 51% Females
- Data split – 80% train, 10% validation, 10% test



Enabling scientific discoveries that improve human health



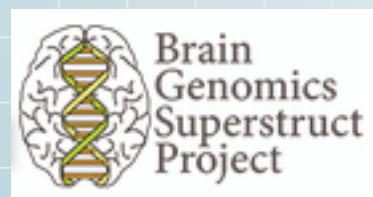
CONSORTIUM FOR  
RELIABILITY AND  
REPRODUCIBILITY



Autism Brain Imaging  
Data Exchange



International Neuroimaging  
Data-Sharing Initiative

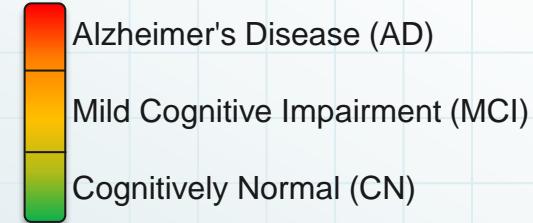
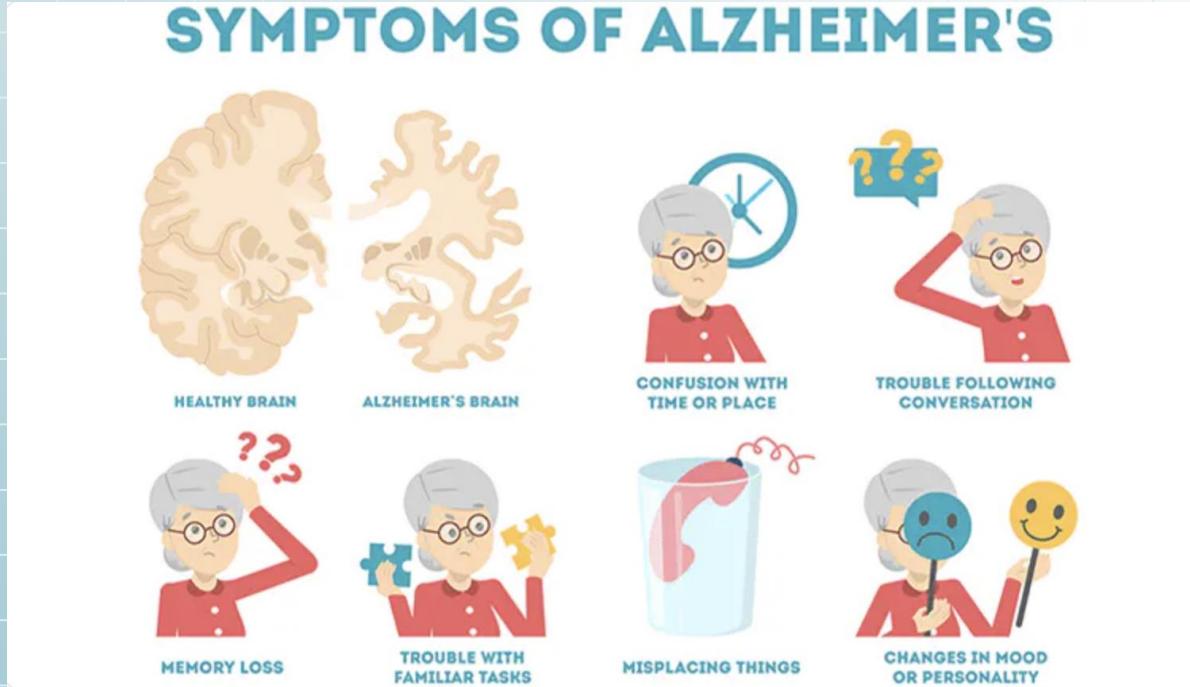


# Results HyperFusion vs Imaging only

- Mean Average Error (MAE) – lower is better

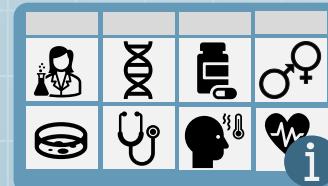
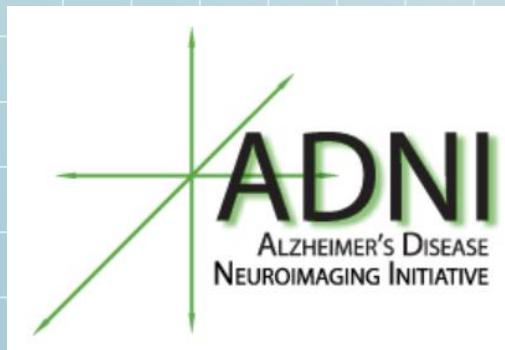


# Alzheimer Disease Classification



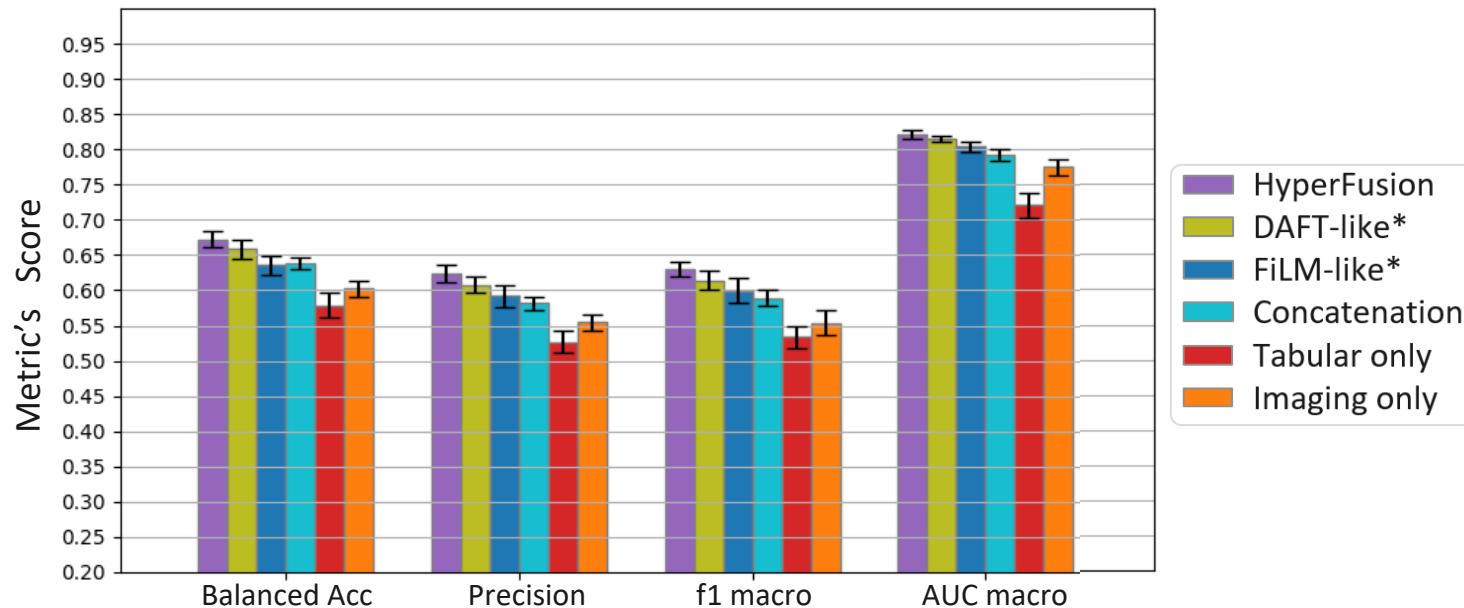
# The Data

- The ADNI dataset, ADNI 1, ADNI 2 and ADNI GO, baseline visits
- ~2120 MRI scans - 34% CN , 48% MCI , 17% AD
- Data split: nested cross validation - 20% test and dynamic 20%-60% validation-train
- The Tabular attributes used (9): demographic, genetic, lab results



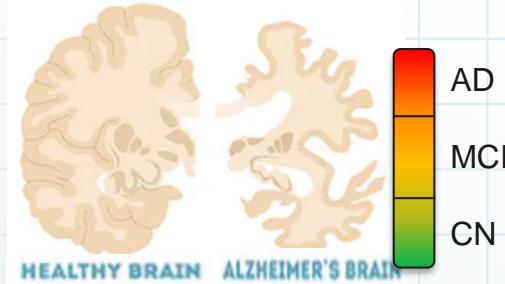
- demographic: Age, Sex, Education (years)
- genetic risk factor: ApoE4
- CSF biomarkers: Abeta42, P-tau181, T-tau
- measures from PET scan: 18F-fluorodeoxyglucose, florbetapir

# Results HyperFusion vs Others



# Summary

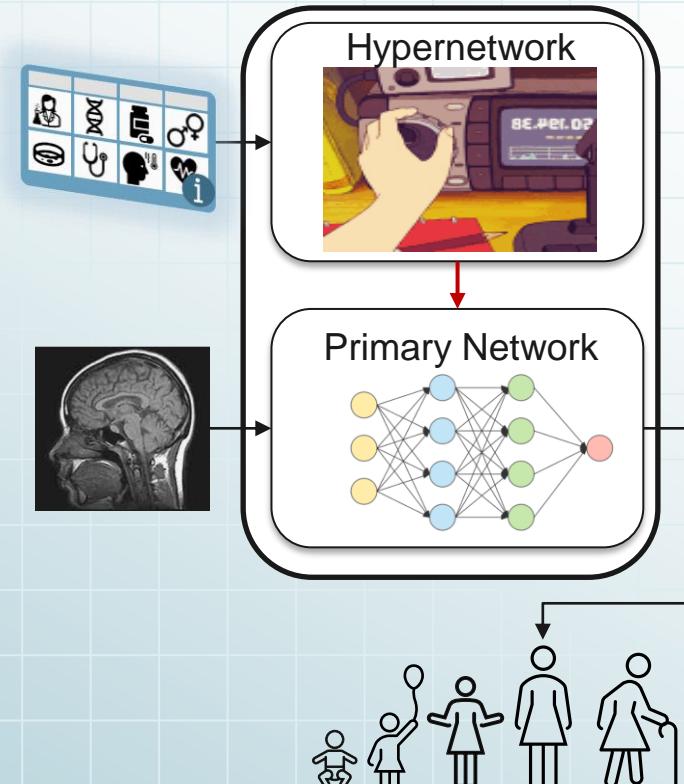
- HyperFusion – a novel hypernetworks approach for multimodal data fusion.
- Key idea – conditioning the image processing by tabular data.
- Versatility – different tasks and architectures
- Outperforms single-modality and existing fusion methods.
- Another step to personalized medical care.



# Thanks for Listening!

**HyperFusion: A Hypernetwork Approach to Multimodal Integration of Tabular and Medical Imaging Data for Predictive Modeling**

Dive into the details on arXiv – scan below



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