

# Discovery of novel biomolecular condensate drug targets in oncology using *in silico* predictive tools

Kadi L. Saar<sup>1</sup>, Marius Rebmann<sup>1</sup>, Mohammed Kanchwala<sup>1</sup>, Seema Qamar<sup>1</sup>, Prathima Radhakrishnan<sup>1</sup>, Jasmine Cornish<sup>1</sup>, Julia Doh<sup>1</sup>, William Arter<sup>1</sup>, Assaf Rotem<sup>1</sup>, Mahmoud Ghandi<sup>1</sup>, Andrew Seeber<sup>1</sup>, Richard C. Centore<sup>1</sup>, Martin Kulander<sup>1</sup>, Tuomas P.J. Knowles<sup>1,2,3</sup>, Sarah A. Teichmann<sup>3,4</sup>, Shilpi Arora<sup>1</sup>

<sup>1</sup>Transition Bio, Cambridge, UK & Watertown, MA, US

<sup>3</sup>Cavendish Laboratory, Department of Physics, University of Cambridge, UK

<sup>4</sup>Wellcome Sanger Institute, Wellcome Genome Campus, Hinxton, Cambridge, UK

<sup>2</sup>Yusuf Hamied Department of Chemistry, University of Cambridge, UK



## TransitionBio

Abstract # 4891

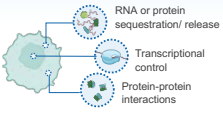
### Intrinsically disordered proteins (IDPs) frequently form biomolecular condensates

#### Opportunity

~70% of cancer-associated proteins are IDPs

#### Challenge

IDPs are difficult to drug with traditional screening methodologies



#### Condensates

IDPs frequently form liquid-like biomolecular assemblies that control macromolecular interactions

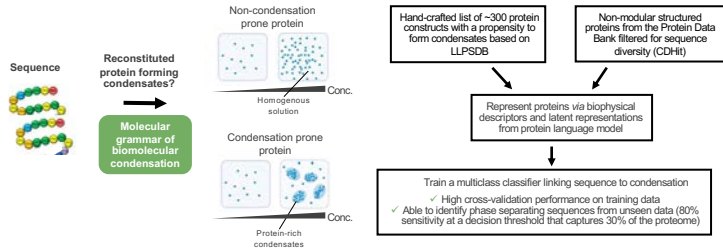
#### Disease causality

Condensates can drive aberrant transcription, translation & protein interactions

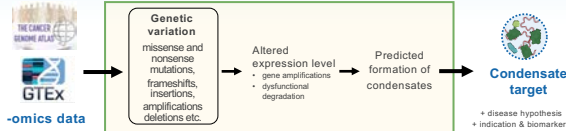
### Use of protein language models and biophysical descriptors to develop predictive machine learning models that link protein's sequence to its condensation behavior

**Task:** Develop a model that links a protein's sequence to its propensity to form condensates

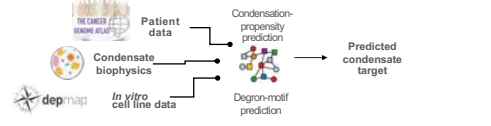
**Training:** Use experimental data, biophysical descriptors and latent representations from pre-trained protein language model to train the model



### Hypothesis: Genetic variations can lead to aberrant condensation via increased intracellular abundance of the target protein



### Identification of gain-of-function mutations in degran domains that can lead to target overexpression and aberrant condensation uncovers $\beta$ -catenin as a target



- Identify degran domain mutations within TCGA proteomics samples<sup>1</sup>
- Use developed ML models to evaluate if protein forms condensates
- Evaluate if *in vitro* cell models show selective dependency on the target<sup>2</sup>
- Analyze if mutations elevate dependency in *in vitro* models or lead to reduced survival in patient samples

*In vitro* CRISPR knock-out dependency data (DepMap) confirms elevated CTNNB1 dependency in the models that carry mutations in the identified driver region

• 108 proteins with confirmed degran motifs

• 73 proteins have a high phase separation propensity

• 41 genes have a strong dependency in *in vitro* models (minimum CRISPR Chronos score < -1)

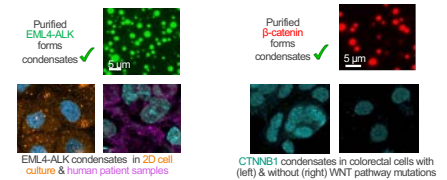
• 26 genes with selective dependency (normality likelihood ratio test<sup>2</sup>)

Propose CTNNB1 ( $\beta$ -catenin) for experimental validation due to high frequency of the mutations within the TCGA samples

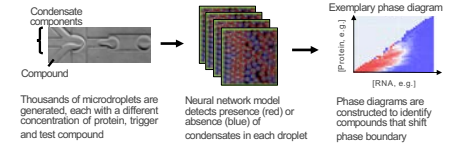
<sup>1</sup> Martinez-Jimenez et al. *Nat Cancer*, 2019  
<sup>2</sup> McDonald III et al. *Cell*, 2017

Predicted targets form condensates and are amenable to screening with Transition Bio's platform

1. *In silico* target ID pipeline predicted targets form condensates in a purified form (top) and in the disease models (bottom)



Transition Bio has established a proprietary microfluidics and machine learning enabled molecular screen for identifying compounds that alter condensates

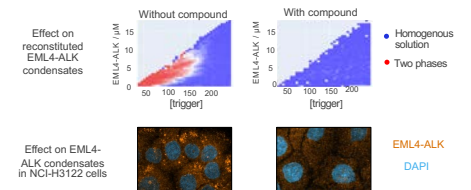


Thousands of microdroplets are generated, each with a different concentration of protein, trigger and test compound

Neural network model detects presence (red) or absence (blue) of condensates in each droplet

Phase diagrams are constructed to identify compounds that shift phase boundary

2. Screen identifies compounds that modulate reconstituted and cellular EML4-ALK condensates



Effect on reconstituted EML4-ALK condensates

Effect on EML4-ALK condensates in NCI-H1272 cells

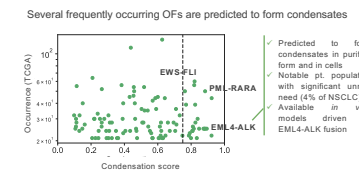
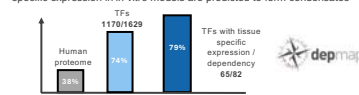
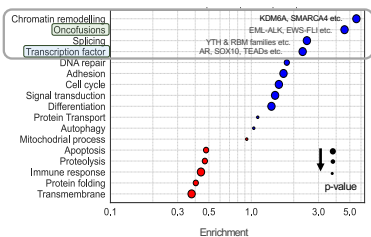
### Summary

- We developed machine learning models that link protein sequence to its propensity to form condensates
- Oncofusions and proteins involved in splicing and transcription are predicted to frequently form condensates via condensate-based mechanisms
- We combined -omics data with our predictive condensation models to identify genetic variations that lead to aberrant condensation
- Predicted targets from *in silico* target identification pipeline (EML4-ALK,  $\beta$ -catenin) form condensates, phase separate *in vitro* and are amenable to screening with our platform

### Proteome-wide *in silico* screen using the developed model predicts oncofusions (OFs) and transcription factors (TFs) as protein classes that frequently form condensates

Oncofusions (OFs), transcription factors (TFs), and proteins associated with chromatin remodeling and splicing have high predicted condensation propensity relative to proteome (enrichment calculated with Fisher test)

72% human TFs and 79% of TFs with selective dependency and/or tissue specific expression in *in vitro* models are predicted to form condensates



TCGA survival data suggests patients with higher level of beta-catenin have a worse survival prognosis (pancreatic adenocarcinoma cohort)

