

Testing the Gene Expression Classification of the EMT Spectrum

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Supplementary Information

1. Construction of the gene regulatory network of EMT

The gene regulatory network of EMT (**Fig. 1A**) is an extended version of the one used in our previous work [1]. The experimental evidence for the added genes and regulatory links are listed in **Table S1** and the experimental evidence for other genes and links can be found in [1].

2. E and M samples purification

RACIPE generates many gene expression samples (each of them being a steady state solution of a given set of parameters and initial conditions) and clusters them into four classes, which can be mapped on to E, hybrid E/M and M phenotypes. In order to isolate RACIPE-generated samples that are highly epithelial, E samples characterized by $CDH1^{high}/CLDN7^{high}/OVOL2^{high}/GRHL2^{high}/\Delta Np63\alpha^{high}$ are selected as purified E samples. These represent a purified subset selected from the distribution of all E samples and represent a signature for a maximally epithelial phenotype. Similarly, purified M samples are identified by selecting those with $VIM^{high}/ZEB1^{high}/ZEB2^{high}/FOXC2^{high}/TGF-\beta^{high}$.

3. Prediction of the mixture proportions for Mixtures of purified E and M samples

In order to assess the ability of the EMT scoring metric to resolve relative proportions of epithelial and mesenchymal signatures from a mixture, we analyzed signatures for mixtures of varying epithelial and mesenchymal proportions. The mean EMT metric-estimated proportions [E, M] in five mixtures are [56%,

44%], [46%, 54%], [65%, 35%], [25%, 75%], [83%, 17%], with respective mean error [5.6%, 6.0%, 5.1%, 4.8%, 3.2%] across replicates, which closely estimate their respective true proportions [50%, 50%], [40%, 60%], [60%, 40%], [20%, 80%], [80%, 20%].

Supplementary Tables

Table S1. Experimental evidence for the regulatory links of OVOL2, GRHL2, Δ Np63 α and CLDN7.

Regulation	References
Mutual inhibition between OVOL2 and ZEB1	[2]
Activation of OVOL2 by GRHL2	[3,4]
Mutual Inhibition between ZEB1 and GRHL2	[5]
Activation of CDH1 by OVOL2	[6]
Self-inhibition of OVOL2	[7]
Inhibition of VIM by OVOL2	[7]
Inhibition of SNAI2 by OVOL2	[7]
Inhibition of Δ Np63 α by OVOL2	[7]
Inhibition of TGF- β by OVOL2	[8]
Inhibition of CLDN7 by ZEB1	[9]
Activation of CDH1 by CLDN7	[10]
Inhibition of SNAI2 by CDH1	[11]
Activation of SNAI2 by Δ Np63 α	[12]
Activation of miR-205 by Δ Np63 α	[12]
Mutual activation between Δ Np63 α and GRHL2	[13]

Table S2. Prediction error for RACIPE-generated gene expression data

A	EMT Metric Predictions: RACIPE Data			
	Category	Binned as E	Binned as E/M	Binned as M
	E	45.06%	54.94%	0.00%
	E/M I	25.69%	74.25%	0.05%
	E/M II	0.21%	95.12%	4.67%
	M	0.02%	28.72%	71.25%
Diagnostic Accuracy: 60.83%				
B	EMT Metric Predictions: Purified RACIPE Data			
	Category	Binned as E	Binned as E/M	Binned as M
	Pure E	98.95%	1.05%	0.00%
	Pure M	0.00%	6.67%	93.33%
	Diagnostic Accuracy: 96.57%			

Table S3. Prediction error for mixtures

A	Absolute Mixture Error: RACIPE Data					
	Category	Mix 1	Mix 2	Mix 3	Mix 4	Mix 5
	Replicate 1	0.002	0.018	0.008	0.006	0.000
	Replicate 2	0.023	0.009	0.001	0.003	0.002
	Replicate 3	0.016	0.006	0.013	0.006	0.000
	Average	0.013	0.011	0.007	0.005	0.001
B	Absolute Mixture Error: Purified RACIPE Data					
	Category	Mix 1	Mix 2	Mix 3	Mix 4	Mix 5
	Replicate 1	0.002	0.000	0.003	0.002	0.001
	Replicate 2	0.001	0.001	0.002	0.003	0.002
	Replicate 3	0.001	0.001	0.006	0.001	0.001
	Average	0.001	0.001	0.004	0.002	0.001

Supplementary Figures:

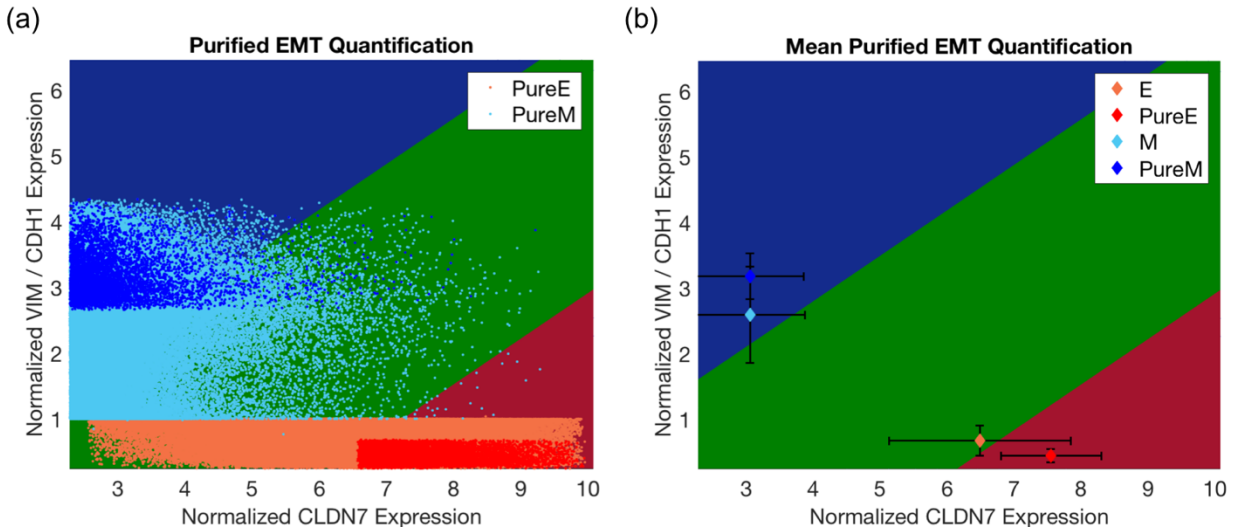


Figure S1. Characterization of the purified E and M samples. (a) Population level projection of RACIPE generated data onto EMT metric predictor space. VIM, CDH1, and CLDN7 expression values for each sample generated from RACIPE are plotted for various populations as categorized by RACIPE (red corresponding to E, green corresponding to hybrid E/M I and hybrid E/M II, and blue corresponding to Mesenchymal) and compared to NCI-60 gene expression training data (black). (b) Individual population members are plotted for each RACIPE-generated category as in Fig. 2A (salmon corresponding to E, light green to hybrid E/M I, magenta to hybrid E/M II, and sky blue to M). Purified E (resp. M) samples are represented by dark red (resp. blue) subsets of their respective populations. Black error bars represent standard deviations along each dimension for all samples in each average.

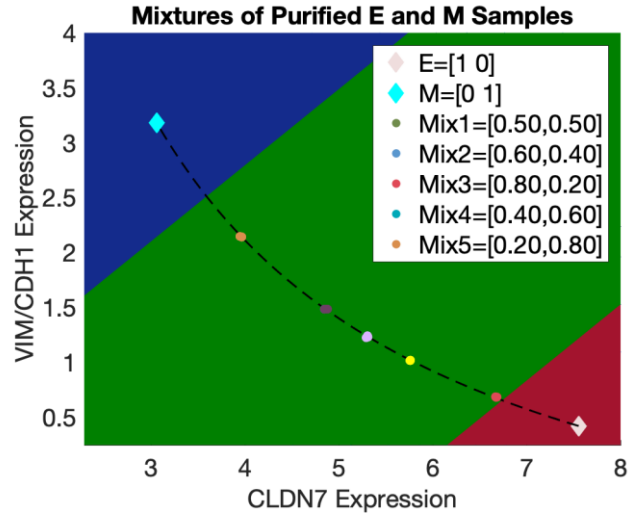


Figure S2. Epithelial and mesenchymal subpopulation estimation by the EMT metric. Predicted mixture proportions contained in each mixture of purified E and M samples are reported in the figure legends and predicted based on their projection to the curve of convex combinations for each non-mixed sample pair. In all cases, all three replicates for one mixture are tightly distributed and therefore have significant overlap.

Reference

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