TCL1A and ATM are co-expressed in chronic lymphocytic leukemia cells without deletion of 11q

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Online Supplementary Design and Methods

Modeling and statistics

Gene expression profiling data were quantile normalized and single genes ranked according to the highest standard deviation between stimulated (murine and human feeder cells) versus unstimulated culture and the highest standard deviation between time points. This identified genes most deregulated upon co-culture compared to unstimulated culture. The 0.5% most deregulated genes (243 of 48600) were tested with Ingenuity Pathway Analysis (IPA, Ingenuity www.ingenuity.com) Systems. and DAVID⁷ (http://david.abcc.ncifcrf.gov) for functional annotation with cell death, apoptosis and cell survival. This identified 35 genes in common between IPA and DAVID, from which the 23 most deregulated genes were selected by visual inspection for mod-

eling in a 2-step approach. First, candidate genes were further grouped into clusters according to their gene expression kinetics using the Partitioning Around Medoids (PAM) clustering method⁸ to yield a number of genes small enough for subsequent network modeling. Here, genes (medoids) were selected from a cluster that displays a gene expression time course kinetic representative of the whole cluster of genes. The resulting 8 medoids were the genes corresponding to the network nodes in the second step of the statistical analysis. Network estimation was made using the dynamic Bayesian network approach.⁹ where probabilities of interdependencies of genes (nodes) are displayed also dependent on kinetics (time). The analysis employs a Markov Chain Monte Carlo algorithm to obtain the posterior edge probabilities of the network. These 8 genes were also used as controls for analysis of transcription factor binding (Figure 2J).

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Online Supplementary Table S1. Primer sequences.

TBP_for:	5`-GCCCGAAACGCCGAATAT-3´
TBP_rev,	5`- CCGTGGTTCGTGGCTCTCT-3´
SDHA_for:	5`-TGGGAACAAGAGGGCATCTG-3´
SDHA_rev:	5°-CCACCACTGCATCAAATTCATG-3′
TCL1A_for:	5`-TCCAGTTTCTGGCGCTTAGT-3´
TCL1A_rev:	5`-ACATCAGTCATCTGGCAGCA-3´
ATM_for:	5`-AGAGGCCGGAAGATGAAACT-3´
ATM_rev:	5`-TGCCTTCTTCCACTCCTTTC-3´







Online Supplementary Figure 2. siRNA knockdown of TCL1a causes 40-80% reduction of transcript levels *MEC2* cells were transfected with 1 g non-target siRNA (*random*) or a siRNA specifically targeting *TCL1A* using Amaxa in two independent experiments. TCL1A expression was measured by qRT-PCR and normalized to *LMNB1* expression.