SUV420H2 is an epigenetic regulator of epithelial/mesenchymal states in pancreatic cancer

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Epithelial-to-mesenchymal transition is implicated in metastasis, where carcinoma cells lose sessile epithelial traits and acquire mesenchymal migratory potential. The mesenchymal state is also associated with cancer stem cells and resistance to chemotherapy. It might therefore be therapeutically beneficial to promote epithelial identity in cancer. Because large-scale cell identity shifts are often orchestrated on an epigenetic level, we screened for candidate epigenetic factors and identified the histone methyltransferase SUV420H2 (KMT5C) as favoring the mesenchymal identity in pancreatic cancer cell lines. Through its repressive mark H4K20me3, SUV420H2 silences several key drivers of the epithelial state. Its knockdown elicited mesenchymal-to-epithelial transition on a molecular and functional level, and cells displayed decreased stemness and increased drug sensitivity. An analysis of human pancreatic cancer biopsies was concordant with these findings, because high levels of SUV420H2 correlated with a loss of epithelial characteristics in progressively invasive cancer. Together, these data indicate that SUV420H2 is an upstream epigenetic regulator of epithelial/mesenchymal state control.

Introduction

The vast majority of cancers originate in epithelial tissues, yet tumors comprise a heterogeneous mix of cell populations with varying phenotypes along the epithelial-mesenchymal continuum (Tam and Weinberg, 2013). Three distinct lines of rationale imply that, in cancer, the epithelial cell state is clinically more favorable than the mesenchymal state. First, normal epithelial cells are stationary, sharing cell–cell junctions and resting on a basement membrane, whereas mesenchymal cells are motile and more likely to migrate and invade (Thiery, 2003). Accordingly, an epithelial-to-mesenchymal transition (EMT) is often thought to accompany the progression of early cancer lesions to invasive malignancies and eventually metastasis (Yang and Weinberg, 2008). Second, the mesenchymal cell fraction in tumors possesses increased “stemness,” including superior capability for self-renewal and differentiation potency, marker expression of tissue stem cells, and elevated tumor-initiating ability (Brabletz et al., 2005; Mani et al., 2008; Rhim et al., 2012; Scheel and Weinberg, 2012). Third, mesenchymal cancer cells universally exhibit lower sensitivity to anticancer drugs than their epithelial counterparts (Yauch et al., 2005; Neve et al., 2006; Witta et al., 2006; Sayan et al., 2009), and malignant cells engage in EMT to acquire drug resistance (Singh et al., 2009; Wilson et al., 2014a,b).

Although the core signaling pathways (TGFβ, NOTCH, WNT, FGF, and BMP) and transcription factors (ZEB1/2, SNAIL, SLUG, TWIST1/2, E47, and FOXC1) that regulate epithelial/mesenchymal cell states have been well characterized (Thiery et al., 2009), attempts at modulating these agents to elicit a mesenchymal-to-epithelial transition (MET) in cancers have been largely unsuccessful in cancer patients (Ginnebaugh et al., 2014). More recently, efforts have focused on manipulating the epigenetic programs that likely govern epithelial/mesenchymal cell states. Although incompletely understood, different classes of histone modifiers have been implicated in these processes in various cancers: the deacetylases HDAC1/2 (Peinado et al., 2004; von Burstin et al., 2009), the demethylases KDM1A (Lim et al., 2010; Lin et al., 2010), PHF2 (Pattabiraman et al., 2016), and LOXL2 (Peinado et al., 2005) and the methyltransferases EZH2 (Cao et al., 2008), EHMT2, and SUV39H1 (Dong et al., 2013). Histone modifiers are attractive targets for prospective therapies because they contain distinct, druggable catalytic domains with some Food and Drug Administration–approved inhibitors already in the clinic and several more in clinical trials (Dawson and Kouzarides, 2012; Jones et al., 2016).

Pancreatic cancer is one of the deadliest malignancies because it is usually detected late in the course of the disease and existing treatments are typically ineffective because of intrinsic and acquired drug resistance, as well as being poorly responsive to immunotherapy (Xiong et al., 2006; Arumugam et al., 2009; Li et al., 2013; Chen and Mellman, 2017). Priming pancreatic cancers with an epithelial-inducing agent might not
only decrease invasion and metastasis and limit stemness but may also increase responses to existing cancer drugs (Singh and Settlemier, 2010). Indeed, histopathological changes associated with pancreatic cancer do not appear to be strictly under genetic control (Lo et al., 2012).

We devised an arrayed screen targeting 300 epigenetic factors and identified SUV420H2 (KMT5C) as an upstream orchestrator of epithelial/mesenchymal states in pancreatic cancer cells. SUV420H2 silences several drivers of MET, and repressing SUV420H2 elicits a molecular, phenotypic, and functional cell identity shift toward the epithelial condition. Analysis of human pancreatic ductal adenocarcinoma (PDAC) samples corroborated a close link between SUV420H2 expression and epithelial/mesenchymal cell states. These findings suggest that SUV420H2 should be considered a potential target to favor MET in pancreatic cancer.

Results

Genetic screen identifies SUV420H2 as a modulator of epithelial/mesenchymal cell states in pancreatic cancer

We designed an unbiased genetic screen to identify and rank epigenetic factors that modulate epithelial/mesenchymal states in pancreatic cancer (Fig. 1 A). The parental PANC-1 cell line, originally derived from the primary tumor of a patient with PDAC with invasion in the duodenal wall and peripancreatic lymph metastasis (Lieber et al., 1975), shows generally poor differentiation, high migration and invasion potential, and marker expression in line with the mesenchymal state (Deer et al., 2010; Klijn et al., 2015). Using fluorescently tagged monoclonal antibodies, we confirmed PANC-1 cells show high levels of the mesenchymal marker vimentin (VIM) and background levels of the epithelial marker E-cadherin (E-CAD) and the epithelial cell adhesion molecule (EPCAM; Fig. S1). In a course of 8 d, PANC-1 cells were subjected to two rounds of transfection by using an arrayed siRNA library targeting 300 genes involved in modulating epigenetic marks in DNA and histones (Table S1). We then subjected the cultures to immunofluorescence using antibodies against VIM, E-CAD, and EPCAM and used a quantitative imaging platform to seek significant changes in average fluorescent intensities per cell.

Each gene was targeted by four different siRNA sequences so we could prioritize genes whose knockdown with at least two siRNAs changed marker signal intensity in a highly significant manner (Fig. 1 B). Although knockdown of numerous genes elicited a change in a single marker and several in two markers, only two epigenetic knockdown targets (SUV420H2 and INTS12) elicited a change in all three phenotypic markers (i.e., VIM, E-CAD, and EPCAM; Fig. 1 C and Fig. S1). INTS12 is a subunit of the snRNA-processing integrator complex and recognizes posttranslational modifications of histone H3 via its PHD finger region (Hernandez, 1985) but has no known enzymatic activity. We therefore concentrated our efforts on SUV420H2 (Suppressor of variegation 4–20 homologue 2), also known as KMT5C, a well-characterized histone methyltransferase that specifically trimethylates Lys-20 of histone H4 causing transcriptional repression of associated genes (Kourmouli et al., 2004; Schotta et al., 2004, 2008). Knockdown of SUV420H2 in PANC-1 cells induced a strong de novo signal of the epithelial markers E-CAD and EPCAM, which correctly localized to cell-cell junctions and substantially diminished signal intensity of VIM (Fig. 1 D).

We confirmed that the SUV420H2 siRNAs used in the screen indeed diminished the expression of their intended target in PANC-1 cells on the RNA and protein level (Fig. S2, A and B). Expression profiling via RNA sequencing (RNaseq) of PANC-1 cells after SUV420H2 knockdown resulted in 266 highly and significantly up-regulated transcripts within arbitrary cutoffs: more than a fourfold change, P < 0.00005 (Table S2). Of those, a large proportion (56 of 266) encoded products that comprise or modulate cell adhesions (e.g., CDH1, EPCAM, CLDN6, and TJP3), cytoskeleton (e.g., various keratins), and extracellular matrix (e.g., various collagens, FNI, and VCAN; Fig. 2 A). Four more genes are specifically expressed in epithelial tissues (ELP3, ESRP1, HNF4A, and RAB17), and another three genes (FOXA1, OVOL1, and OVOL2) encode transcription factors previously implicated in MET (Song et al., 2010; Roca et al., 2013). Unbiased gene set enrichment analyses by Camera (Wu and Smyth, 2012) yielded “Hallmark epithelial-mesenchymal transition” as a significantly altered set among the Broad’s MsigDB Hallmark gene sets (Liberzon et al., 2015; Table S3), as well as “Reactive cell–cell junction organization” as one of the top altered signatures from MsigDB’s C2 curated signature collection (Liberzon et al., 2011; Table S4).

RNaseq data, as well as quantitative RT-PCR (qRT-PCR), corroborated the findings of the genetic screen regarding CDH1 (E-CAD), EPCAM, and VIM on SUV420H2 knockdown; CDH1 and EPCAM were markedly up-regulated (11.1–14.3-fold and 5.2–5.5-fold, respectively), and VIM’s expression level was decreased to 0.5–0.7-fold the original values (Fig. 2 B).

Aside from VIM, RNaseq showed quantitatively variable (10–80%) but statistically significant down-regulation of other mesenchymal markers (CDH2, CDH11, and MMP1) and EMT factors (ZEB1, E47, FOXI1, and FOXC2; Fig. 2 C). SNAI1 expression was unaltered, whereas other EMT transcription factors (ZEB2, SNAI2/3, and TWIST1/2) were expressed at undetectable or negligible levels in parental PANC-1 cells and remained so after SUV420H2 knockdown. Conversely, there was statistically significant up-regulation (up to 25-fold) of various genes implicated in MET (KLF4, BMP1, BMP5, BMP8A, PKCA, and PKCB and the previously mentioned FOXA1, OVOL1, and OVOL2; Fig. 2 D).

The observed molecular transition was reversible. When PANC-1 cells were subjected to the 8-d SUV420H2 knockdown regimen followed by 8 d of recovery, expression levels of CDH1, EPCAM, and VIM largely reverted to their original levels (Fig. S2 C).

Knockdown of SUV420H2 in four other pancreatic cancer cell lines with mesenchymal identity (Klijn et al., 2015) followed by qRT-PCR for CDH1, EPCAM, and VIM indicated that aspects of MET were recapitulated in all of them (Fig. 2 E). Of those, the three cell lines with the most robust changes on the RNA level were subjected to Western blotting, showing that E-CAD and VIM were up- and down-regulated, respectively, on a protein level (Fig. 2 F). When instead SUV420H2 was knocked down in mesenchymal cancer cell lines derived from other tissues (Klijn et al., 2015), only occasional alterations of CDH1, EPCAM, and VIM expression levels were observed (Fig. S2 D).

Collectively, these results indicated that SUV420H2 is a potent epigenetic factor in controlling epithelial/mesenchymal identity status in pancreatic cancer cells, and its repression elicits a global MET from a molecular standpoint.
SUV420H2 knockdown reduces migration, invasion, stemness, and chemoresistance in pancreatic cancer cells

We next investigated the phenotypic and functional changes in pancreatic cancer cells after SUV420H2 knockdown. Mesenchymal cells that delaminate from cellular masses are more motile compared with their epithelial counterparts, which are typically static or move en masse (Nieto, 2013). Subconfluent parental PANC-1 cells in monolayer typically grow in loose assemblies, where the bulk of cells groups in clusters and several single cells intersperse in the spaces between those clusters (Fig. 3A). When SUV420H2 was knocked down, PANC-1...
Figure 2. Pancreatic cancer cells undergo molecular MET on knockdown of SUV420H2. (A) Volcano plot of RNaseq results from SUV420H2 knockdown in PANC-1 cells, compared with siNTC control. Listed genes fall within the arbitrary region of interest (dashed box), defined by a fold change >4 and P < 0.00005. (B) Expression analysis of CDH1 (gene encoding E-CAD), EPCAM, and VIM on SUV420H2 knockdown in PANC-1 cells by RNaseq and qRT-PCR. Bar graphs indicate mean ± SD. For RNaseq, n = 3 biological replicates and differences assessed by using voom+limma (see Materials and methods). For qRT-PCR, n = 3 biological replicates each averaged from three technical replicates; differences were assessed by Student’s t test compared with siNTC control. ***, P < 0.001; ****, P < 0.0001. (C) Expression analysis from RNaseq for mesenchymal markers and EMT-inducing factors in PANC-1 cells with SUV420H2 knockdown. Data are represented as normalized to the siNTC control knockdown. Bar graphs indicate mean ± SD. n = 3 biological replicates and differences assessed by using voom+limma. *, P < 0.05; ****, P < 0.0001. (D) Expression analysis from RNaseq for MET-inducing factors in PANC-1 cells with SUV420H2 knockdown. Data are represented as normalized to the siNTC control knockdown. Bar graphs indicate mean ± SD. n = 3 biological replicates and differences assessed by using voom+limma. ***, P < 0.001; ****, P < 0.0001. (E) qRT-PCR expression analysis for CHD1, EPCAM, and VIM on SUV420H2 knockdown in pancreatic cancer cell lines of mesenchymal identity. Data are represented as average fold change normalized to the siNTC control knockdown. Maxima for CDH1, EPCAM, and VIM are 16.47-, 5.50-, and 0.42-fold, respectively. Heat map data depicts means. n = 3 biological replicates each averaged from three technical replicates. (F) Western blots for E-CAD and VIM in mesenchymal pancreatic cancer cell lines with control and SUV420H2 knockdown. Loading control is a-TUBULIN (α-TUB).
cells grew in compact cohorts with clearly defined margins (Fig. 3A). Scratch-wound and Matrigel-based transwell assays showed a clear reduction of migration and invasion potential of three mesenchymal pancreatic cancer cell lines on SUV420H2 knockdown (Fig. 3, B and C; and Fig. S3, A and B).

Within the bulk population of tumors, the subset of cells with mesenchymal traits has been correlated with stemness in several cancer types (Scheel and Weinberg, 2012). Pancreatic cancer cells with the profile CD24hi CD44hi are often referred to as “cancer stem cells” because of their superior ability to form tumors in xenografts and increased efficiency at forming tumorspheres, an often-used readout of stemness (Huang et al., 2008; Yin et al., 2011). On knockdown of SUV420H2, mesenchymal pancreatic cancer cells exhibited significantly decreased expression of CD24 and CD44 (Fig. 4, A and B) and had decreased potential to generate pancreatic tumorspheres (Fig. 4 C).

Finally, mesenchymal cancer cells are universally more resistant to anticancer drugs than their epithelial counterparts (Singh and Settleman, 2010). SUV420H2 knockdown rendered mesenchymal pancreatic cancer cells significantly more sensitive to gemcitabine and 5-fluorouracil, two of the most
commonly used chemotherapies in human PDAC (Ghadban et al., 2017; Fig. 4 D and Fig. S3 C).

Collectively, those data implied a shift from a mesenchymal to an epithelial state in pancreatic cancer cells at both the phenotypic and functional levels as a result of SUV420H2 knockdown.

We next asked if SUV420H2 might act by selectively modulating transcription factors associated with MET. Our RNaseq analysis demonstrated that on SUV420H2 knockdown the transcription factors FOXA1, OVOL1, and OVOL2 became highly

Figure 4. SUV420H2 knockdown decreases stemness and anticancer drug resistance of mesenchymal pancreatic cancer cells. (A) CD24 and CD44 RNA expression levels with NTC control or SUV420H2 knockdown determined by RNaseq or qRT-PCR. For graphs, bars indicate mean ± SD. For RNaseq, n = 3 biological replicates and differences assessed by using voom+limma (see Materials and methods). For qRT-PCR, n = 3 biological replicates each averaged from three technical replicates, and differences were assessed by Student’s t test compared with siNTC control. *, P < 0.05; **, P < 0.01; ***, P < 0.001; ****, P < 0.0001. (B) Cell-surface protein level for CD24 and CD44 analyzed by flow cytometry in PANC-1 cells. B shows one representative panel per condition of a set of triplicates; differences were assessed by Student’s t test compared with siNTC control. ***, P < 0.001; ****, P < 0.0001. (C) Tumorsphere formation assay showing representative images from a PANC-1 experiment and quantitation in three pancreatic cancer cell lines. n = 60 fields of view considered for each cell line, composed of 20 across three biological triplicates. Bar graphs depict mean ± SD. Differences were assessed by Student’s t test compared with siNTC control in each cell line. *, P < 0.05; ****, P < 0.0001. Bars, 250 µm. (D) Cell viability assays demonstrating effects of Gemcitabine and 5-Fluorouracil (5-FU) over 72 h on mesenchymal pancreatic cancer cells with NTC control or SUV420H2 knockdown, with two siRNAs for each. n = 6 biological replicates. Significance for each drug concentration was assessed by two-way analysis of variance compared with siNTC #1 control and indicated as pink and orange asterisks for siSUV420H2 #1 and siSUV420H2 #2, respectively. IC50 for each condition is displayed in the table. *, P < 0.05; **, P < 0.01; ***, P < 0.001; ****, P < 0.0001.

SUV420H2 regulates several epithelial-promoting transcription factors

We next asked if SUV420H2 might act by selectively modulating transcription factors associated with MET. Our RNaseq analysis demonstrated that on SUV420H2 knockdown the transcription factors FOXA1, OVOL1, and OVOL2 became highly
up-regulated in PANC-1 cells, and we observed increased levels of FOXA1 by Western blots in various mesenchymal pancreatic cancer cells (Fig. 2, A and D; and Fig. S3 D). Previously published observations in pancreatic cancer cells showed that expression of FOXA1 was sufficient to neutralize several E-CAD repressive mechanisms and that its knockdown induced EMT (Song et al., 2010). Other studies showed that OVOL1/2 expression facilitated MET in several cancer types by repressing ZEB1 and inducing ESRP1, which regulates RNA splicing to generate epithelial-specific isoforms (Roca et al., 2013).

Consequently, we reasoned that if SUV420H2 was acting via one or more of these transcription factors, double-knockdown experiments should neutralize the MET-inducing effects of SUV420H2 knockdown alone. Indeed, double knockdowns in PAN−1 cells of SUV420H and FOXA1, OVOL1, or OVOL2 prevented key molecular features of MET (Fig. 5 A). In particular, the expression of the epithelial genes CDH1 and EPCAM was significantly reduced in all cases, even though expression of the mesenchymal gene VIM was unaffected. ZEB1 expression, which was significantly reduced in PAN−1 cells on SUV420H2 knockdown, was partially rescued when FOXA1, OVOL1, or OVOL2 was also knocked down (Fig. S3 E). Therefore, these double-knockdown “rescue” experiments argued that SUV420H2 affected CDH1, EPCAM, and ZEB1 via all three tested MET inducers.

To confirm that SUV420H2 controls shifts in the epithelial/mesenchymal state via histone marks, we investigated global patterns of H4K20 histone methylation marks in PAN−1 cells by histone mass spectrometry. Knockdown of SUV420H2 with two different siRNAs was effective at reducing methyltransferase activity and resulted in 29–50% overall global decreases in H4K20me3, the repressive mark produced by SUV420H2 (Fig. 5 B). Conversely, there was a global increase in the unmethylated and monomethylated states of H4K20, both of which have been associated with an activated transcriptional state for a variety of genes (Karachentsev et al., 2005; Talasz et al., 2005; Li et al., 2011). Western blots in PAN−1, as well as two other mesenchymal pancreatic cancer lines (Hs 766T and KP4), showed a reduction in the H4K20me3 mark on SUV420H2 knockdown (Fig. 5 C).

We went on to analyze the status of the H4K20me3 mark specifically on FOXA1, OVOL1, and OVOL2 by chromatin immunoprecipitation coupled with quantitative PCR (ChIP-qPCR; Fig. 5 D and Fig. S3 F). Being a known repressive histone mark, trimethylation of H4K20 around their respective genes could be the mechanism through which SUV420H2 controlled expression of these three transcription factors. Histone mark profiles can vary across the sequence of a gene; consequently, we adopted a ChIP-qPCR protocol probing multiple regions per gene in our experiments (Milne et al., 2009). SUV420H2 knockdown in PAN−1 cells resulted in significant reductions in signal in at least one of three loci in each gene tested (Fig. 5 D). This suggested that SUV420H2 controlled the H4K20me3 mark on each of our three candidate MET transcription factors and therefore their capacity to be transcribed. In contrast, SUV420H knockdown had no direct effect on H4K20me3 levels on CDH1 or EPCAM, suggesting that SUV420H2 acts indirectly on these epithelial markers (Fig. S3 G). H4K20me3 levels at the ZEB1 locus were also unaltered, indicating no direct activity of SUV420H2 in controlling the expression of ZEB1 (Fig. S3 G).

Together, these results suggested that SUV420H2 controls the dynamics of H4K20me3 marks on specific genes encoding transcription factors that control MET.

## SUV420H2 gain of function induces EMT in pancreatic cancer cells

To test whether SUV420H2 gain of function could elicit the opposite effect we observed with its knockdown, we chose two epithelial pancreatic cancer cell lines (Klijn et al., 2015) and induced expression of a wild-type SUV420H2 and a methyltransferase dead allele of SUV420H2 using overexpression vectors. For the latter, we designed a sequence with an in-frame 42-bp deletion, resulting in a 14–amino acid excision within the catalytic domain of SUV420H2. We confirmed that the overexpression constructs could induce expression of the SUV420H2 alleles (Fig. 6 A) and that the wild-type allele, but not the kinase-dead version, resulted in increased H4K20me3 marks in transfected cells (Fig. 6 B). qRT-PCR showed that epithelial Capan-1 and Panc 04.03 cell lines strongly repressed the epithelial markers CDH1 and EPCAM and induced VIM on SUV420H2 overexpression (Fig. 6 C and Fig. S3 H). In addition, in both cell lines there was significant reduction in FOXA1, OVOL1, and OVOL2, the three MET transcription factors we had previously mechanistically implicated with SUV420H2 (Fig. 6 C and Fig. S3 H). Western blots for E-CAD, VIM, and FOXA1 confirmed these trends on a protein level in both epithelial cell lines (Fig. 6 D).

The observed changes were only observed after overexpression of the wild-type allele of SUV420H2 and not with the methyltransferase dead allele, directly implicating the catalytic activity of SUV420H2 in the transitions in epithelial/mesenchymal cell identity in pancreatic cancer cells.

Together, these experiments showed that up-regulation of SUV420H2 can induce EMT in epithelial pancreatic cancer cells because of its catalytic activity and ensuing H4K20me3 marks.

## Pancreatic adenocarcinoma progression correlates expression of SUV420H2 with EMT

To determine if the regulatory activities of SUV420H2 in pancreatic cancer cell lines correlated with any features of clinical human PDAC, we examined the expression and localization of SUV420H2 and epithelial/mesenchymal markers in archival tumor samples (clinical details on analyzed samples outlined in Table S5). Using the same three antibodies for E-CAD, EPCAM, and VIM as we used for the original in vitro screen, we performed immunofluorescence on sections of PDAC samples, which contained regions of healthy exocrine tissue, early cancer lesions, as well as domains of advance invasive cancer (Fig. 7 A). In healthy exocrine epithelia, E-CAD and EPCAM showed strong signal at cell–cell junctions, and little or no VIM staining was observed (Fig. 7 B and D; and Fig. S4). In pancreatic intraepithelial neoplasia (PanIN) regions, levels of E-CAD and EPCAM were still relatively robust at cell–cell junctions between neighboring epithelial cells (Fig. 7 B and D; and Fig. S4). As expected, the individual epithelial cells within the PanIN lesions appeared to gradually lose their canonical epithelial morphology. PanIN epithelial cells, like healthy epithelia, were negative for VIM.

In invasive carcinoma regions (Fig. 7 B and D; and Fig. S4), epithelial cells displayed a decidedly dysmorphic cell shape and showed weak E-CAD signal and no EPCAM signal. We could not detect any malignant cells that unequivocally displayed de novo VIM expression. VIM levels were strong in stromal cells at all stages analyzed. These observations indicate that, concomitant with pancreatic cancer progression, cells lost epithelial marker expression and morphology and exhibited a
phenotype consistent with one that is more mesenchymal in nature, although they were devoid of VIM staining.

The fact that we did not observe cancer cells clearly positive for VIM could have several explanations. First, in the samples analyzed, cells might not have made a full transition to the mesenchymal state but rather displayed a partial EMT that might be sufficient to delaminate from the normal tissue epithelium. Alternatively, a full transition does happen, but because of the sporadic and asynchronous nature of the transition, it is difficult to capture in static images. Finally, VIM might be expressed in cancer cells but at levels that are undetectable when imaged adjacent to stromal cells, which exhibit strong VIM staining. Nevertheless, we detected a gradual increase in nuclear levels of ZEB1 in progressively invasive pancreatic cancer cells, which indeed again supported the notion that an EMT accompanies PDAC progression (Fig. S5A).

Strikingly, immunofluorescence for SUV420H2 in human PDAC resulted in background/low signal in residual healthy exocrine, but moderate signal in PanIN cells, and strong, nuclear signal in invasive carcinoma cells, showing a clear and gradual increase of SUV420H2 protein levels in parallel with tumor promotion (Fig. 7, C and D; and Fig. S4). Together with the aforementioned findings, this described a direct correlation between pancreatic cancer development, increasing amounts of
SUV420H2, and a shift away from the epithelial and toward the mesenchymal cell state.

Finally, we queried public databases for SUV420H2 transcript levels in human cancers. Data compiled from the largest PDAC study in cBioPortal (Cerami et al., 2012; Gao et al., 2013) suggests that 18% of pancreatic cancers have amplifications of SUV420H2, and records derived from The Cancer Genome Atlas (TCGA; http://cancergenome.nih.gov/) as well as internal Genentech datasets show significantly elevated levels of SUV420H2 in a vast array of cancer types, including bladder, breast, colon, kidney, lung, liver, prostate, stomach, uterus, and others (Fig. S5 B).

Collectively, the data from pancreatic cancer showed an association between EMT and elevated expression of SUV420H2 during the pathological progression of PDAC, and data from other cancers show a global correlation of SUV420H2 with malignancies in a vast array of tissues.

**Discussion**

Our data suggest that SUV420H2 acts as an epigenetic orchestrator of epithelial/mesenchymal cell states in pancreatic cancer. As summarized in a working model in Fig. S5 C, SUV420H2, via its repressive mark H4K20me3, silences expression of the key MET-promoting transcription factors FOXA1, OVOL1, and OVOL2 (Fig. S5 C, left). When SUV420H2 expression is attenuated, its repressive mark disappears, thereby enabling expression of FOXA1, OVOL1, and OVOL2 and activating the epithelial transcriptional program (Fig. S5 C, right). The three MET-promoting transcription factors are likely to directly induce de novo transcription of structural epithelial markers. This has been shown for FOXA1, which binds the regulatory region of CDH1 in pancreatic cancer to control its expression (Song et al., 2010). The activation of the epithelial-like MET program also includes attenuation of EMT factors such as ZEB1, whose transcription was previously shown to be repressed directly by OVOL1/2 via direct promoter binding (Roca et al., 2013). The same study found that OVOL1/2 induced ESRP1, leading to processing of epithelial-specific splice variants. Alternative splicing mediated by ESRPs is a hallmark of epithelial differentiation (Warzecha and Carstens, 2012; Bebee et al., 2015). In agreement with this interpretation, our expression-profiling experiment revealed a sharp increase in ESRP1 levels on knockdown of SUV420H2 (Fig. 2 A).

It remains unclear how SUV420H2 controls expression of the mesenchymal marker VIM. A previous study showed that OVOL1/2 represses VIM (Lee et al., 2014), but in our double knockdown experiments, VIM expression was not rescued by simultaneous siRNA knockdown of SUV420H2, FOXA1, or either of the OVOL genes. Thus, VIM expression appears to be independent of the three SUV420H2-controlled MET transcription factors in our context. It is unlikely that SUV420H2 acts directly on the VIM locus to control its expression, considering H4K20me3 is a repressive mark, and there is a positive correlation between the two factors. Instead, an as yet unidentified parallel signal is likely to play a role. It is also worth noting that, although SUV420H2 knockdown in PANC-1 cells induced robust expression of the epithelial markers, VIM levels were only partially reduced but never completely eliminated, suggesting that cells had achieved a hybrid or transitional state. Conceivably, a MET program was induced but did not go to completion, a transition referred to as “partial MET” (Jolly et al., 2015). Full MET might require other players or take longer time to achieve. Nonetheless, the acquisition of epithelial characteristics was sufficient to reduce cell invasion and migration and increase sensitivity to anticancer drugs.

Although SUV420H2 orchestrates the induction of the transcriptional program controlling MET, a key remaining question is what controls expression of SUV420H2 itself. In pancreatic cancer, the gradual increase in SUV420H2 levels as
the disease progresses must reflect a response to a broader biological mechanism. Because only 18% of PDAC occurrences contain amplifications of \( \text{SUV420H2} \) (Cerami et al., 2012; Gao et al., 2013), amplification does not seem likely to explain the elevated levels of \( \text{SUV420H2} \) in the majority of PDAC cases. Thus, amplification-independent mechanisms leading to over-
expression are likely to be more common. In general, the processes involved in the induction of MET have not been well characterized (Nieto, 2013), so the identification of the signals responsible for enhancing SUV420H2 expression will likely shed light on the conditions that control mesenchymal and epithelial balance in pancreatic cancer.

The analysis of TCGA data on SUV420H2 expression in different cancer types emphatically showed near-universal, statistically significant elevation in malignancies compared with matched healthy tissues (Fig. S5 B). Interestingly, previous studies have generally reported a decrease in H4K20 trimethylation during the tumorigenic process in many cancer types, to the extent that global loss of H4K20 trimethylation was proposed to be a common hallmark of many human tumor cells (Fraga et al., 2005). One fact that might reconcile these observations is that, even though the levels of SUV420H2 might increase in cancers, concomitant high expression of demethylases might outcompete the methyltransferase activity of SUV420H2, resulting in overall lower H4K20me3 levels. This does not appear to be the case in pancreatic cancer, however, and to our knowledge a decrease in H4K20me3 has never been reported for pancreatic malignancies.

It should also be noted that the data reporting a decrease in H4K20me3 in cancers have primarily focused on H4K20me3 marks on repetitive sequences of the genome, particularly in centromeric repeats. Consequently, it is possible that from a global cancer cell standpoint that levels of H4K20me3 decrease, but marks in particular loci such as the MET-transcription factors should increase. A regulator complex that guides SUV420H2 to specific genomic sites could achieve this. A precedent for such a mechanism has been reported, in which SUV420H2 associates with a long noncoding RNA that guides the methyltransferase to specific genes (Bierhoff et al., 2014).

It is also important to note that, even though SUV420H2 transcript levels appear elevated in virtually all types of malignancies, qRT-PCR analysis of epithelial/mesenchymal markers in cancer cell lines from various tissues other than the pancreas rarely resulted in significant changes upon SUV420H2 knockdown (Fig. S2 D). Published work in breast cancer cells has even shown an opposite effect of SUV420H2 regarding epithelial/mesenchymal cell identity, actually rendering cells more epithelial and less invasive (Yokoyama et al., 2014; Shinchi et al., 2015). This would argue for context-dependent activity of SUV420H2 and its mark H4K20me3. In some cancers, there appears to be mechanistic disconnect between SUV420H2 and epithelial/mesenchymal cell identity. Future work might elucidate new functions of SUV420H2 in other cancers, possibly with new targets for methyltransferase activity or scaffolding functions for protein complexes that might help explain its elevated levels across cancer types.

Of the histone modifiers previously associated with the control of epithelial/mesenchymal cell states, their suggested mechanisms of action range from forming complexes with EMT transcription factors to directly silencing or activating expression of EMT drivers or genes encoding structural epithelial traits such as E-CAD (Peinado et al., 2005; Tam and Weinberg, 2013; Pattabiraman et al., 2016). Here, we show an alternative mechanism, namely a histone modifier silencing transcription factors that steer the MET process. From our in vitro cell work we observe that SUV420H2, being a repressor of transcription, seemingly has no direct effect on the classical EMT factors (ZEB1/2, SNAI1/2, and TWIST1/2) but instead selectively silences FOXA1 and OVOL1/2 thereby inhibiting a cell’s ability to remain in the epithelial state. In human PDAC samples, we observed a close correlation between SUV420H2 and loss of epithelial characteristics in cancer progression, which is consistent with previous work showing diminishing levels of FOXA1 with progressive pancreatic cancer (Song et al., 2010). Whether SUV240H2 also plays a broader role in development, particularly in orchestrating MET in pancreatic organogenesis and other tissues, remains to be explored. In the mouse, the SUV420H2 knockout is normal and viable, although the H4K20me3 mark is only partially abrogated in embryonic and adult tissues, suggesting the presence of compensatory methyltransferases that may be active in development but not in cancer (Schotta et al., 2008). In the context of cancer, SUV420H2 could serve as a biomarker of invasive PDAC. It is still unclear whether promoting a MET is in itself a beneficial strategy in oncology, considering that circulating tumor cells that have gone through an EMT at the primary lesion are thought to undergo MET for at the metastatic site (Tsai and Yang, 2013). Consequently, forcing a MET could conceivably facilitate metastasis outgrowth. Nonetheless, promoting the epithelial state by targeting SUV420H2 in combination with conventional chemotherapies and decreasing resistance might prove to be an effective treatment for the devastating diagnosis of pancreatic cancer.

Materials and methods

siRNA screen to identify epigenetic regulators of epithelial/mesenchymal states

PANC-1 cells (CRL-1469; ATCC) were plated at 1,000 cells per well in 96-well plates (CLS3991; Corning) prestampressed in array format with the EP1300 siRNA library composed of siGENOME sequences (Dharmacon) targeting 200 genes involved in modulating epigenetic marks (see the full list of genes, siRNA sequences, and catalog numbers in Table S1). Four different siRNAs per gene were reverse transfected at 12.5 nM with DharmaFECT1 (Dharmacon) in duplicate. Each plate contained siNTC nontargeting controls (D-001810-10; Dharmacon), positive controls (siZEB1; D-006564-01; Dharmacon), and transfection controls (siControl Top; D-001500-01; Dharmacon). Four days later, a second round of transfection was performed concomitant with a media change by using the same library and control layout. After four more days, cells were fixed with 4% paraformaldehyde, washed in PBS, and exposed to the nuclear stain Hoechst 33342 (H3570; Molecular Probes), and the following fluorescently conjugated monoclonal antibodies: rabbit Vim_488 (9854; Cell Signaling Technology), mouse E-Cadherin_555 (560064; BD), and mouse EPC_a1C_488 (52755; BD). Images were acquired with an IN Cell 6000 (GE Healthcare) care at 20x lens magnification with 0.75 NA, 9 images per well, and fluorescent signals quantified with IN Cell imaging analysis software (GE Healthcare). Replicate averaging, intra- and interplate normalization, and subsequent analysis of data and standard (Z) score calculations were performed with ScreenSifter (Kumar et al., 2013). For hit selection, we excluded siRNA sequences that substantially prevented proliferation or caused high levels of cell death (Z score of cell count less than −11.3).

RNaseq

RNaseq was performed by using PANC-1 cells with the following two knockdown conditions: siNTC (nontargeting control) and siSUV420H2
(using sequence 1; Table S1) in triplicates. Total RNA was extracted by using the QIAGEN RNeasy kit per manufacturer’s protocol including the on-column DNase digestion. Quality control of samples was done to determine RNA quantity and quality before their processing by RNaseq. The concentration of total RNA samples was determined by using NanoDrop 8000 (Thermo Fisher). The integrity of RNA samples was determined by using 2100 Bioanalyzer (Agilent Technologies). Approximately 500 ng of total RNA was used as an input for library preparation by using TrueSeq RNA Sample Preparation kit v2 (Illumina). The size of the libraries was confirmed by using 2200 TapeStation and High Sensitivity D1K screen tape (Agilent Technologies), and their concentration was determined by the qPCR-based method by using Library quantification kit (KAPA). The libraries were multiplexed and then sequenced on Illumina HiSeq2500 (Illumina) to generate 30M of single-end 100-bp reads.

RNaseq data processing and analysis
HTSeqGenie (Gregoretti and Reeder, 2014) was used to perform filtering, alignment, and feature counting. HTSeqGenie used GSNAP (Wu and Nacu, 2010) to align reads to the human genome build hg19. Gene models were used from Genentech’s internal database based on Refseq. Gene expression was quantified by counting the number of reads mapping uniquely to any exon of a transcript from the gene. A prefiltering was done for low expression, and only genes with ≥10 read counts in at least three samples (irrespective of condition) were considered. Differential gene expression analysis was performed by using voom+limma (Law et al., 2014). The full RNaseq data discussed in this publication have been deposited in National Center for Biotechnology Information’s Gene Expression Omnibus (Edgar et al., 2002) and are accessible through GEO series accession no. GSE104542.

Gene set enrichment analysis
Camera (Wu and Smyth, 2012) was used to perform competitive gene set testing by using its parametric test. Gene set collections were used from Molecular Signatures Database (M SigDB), including the C2 collection (Liberzon et al., 2011) of canonical pathways and experimental signatures and Hallmark gene sets (Liberzon et al., 2015), which represent coherent genes for well-curated biological processes.

qRT-PCR
qRT-PCR was performed in biological triplicates, each with technical triplicates, on a QuantStudio 5 System (Thermo Fisher) after RNA isolation using the RNeasy kit (Qiagen). cDNA synthesis with SuperScript VILO (Thermo Fisher) and, by using the following Taqman (Thermo Fisher) assays: CDH1 (Hs01023894_m1), EPCAM (Hs00901885_m1), VIM (Hs00185584_m1), SUVA240H2 (Hs00261961_m1 or Hs00938285_g1), CD24 (Hs02379687_s1), CD44 (Hs01075864_m1), ZEB1 (Hs01566410_m1), OVOL (Hs00190060_m1), OVOL2 (Hs01067398_m1), FOXA1 (Hs04187555_m1), and TFRC (Hs00999911_m1), which was used as the normalizer. All cell lines used in qRT-PCR experiments were obtained from Genentech’s in-house cell line repository.

Immunoblotting
Cell lysates were extracted by boiling in 6 M Urea buffer (6 M Urea, 20 mM Tris, pH 7.5, 12.5 mM NaCl, 2.5 mM MgCl2, and 0.1% Triton X-100), supplemented with Halt Protease and Phosphatase Inhibitor cocktail. For histone and histone modifications, extraction was performed by using the EpitQuick nuclear extraction kit (Epigentek) following the manufacturer’s protocol. Protein quantification was performed by using a Pierce BCA Protein Assay kit, and equal amounts of protein lysate were loaded onto the NuPAGE Bis-Tris Gel. Immunodetection and quantification of proteins were performed by a quantitative Western blot method by using LI-COR Odyssey. Antibodies were: mouse E-CAD (Cell Signaling Technology 5296), mouse VIM (3390; Cell Signaling Technology), rabbit FOXA1 (58613; Cell Signaling Technology), rabbit α-TUBULIN (2144; Cell Signaling Technology), mouse H4K20me3 (39671; Active Motif), and rabbit histone H4 (39269; Active Motif).

Cell migration and invasion assays
Cell migration was examined by using the IncuCyte’s (Essen BioScience) ZOOM System and Cell migration kit. Cells were plated at 2,000 cells per well in 96-well ImageLock plates and exposed to a double siRNA transfection course at days 0 and 4 of plating as described for the screen. The conditions were siNTC (nontargeting control) and siSUVA240H2 (using sequence 1 or sequence 2; Table S1), each in six replicates. At the fifth day, the cells were confluent, and a scratch wound was inflicted with a 96-pin woundmaker. Cells were subsequently imaged in an IncuCyte Live-Cell System, and results were analyzed with IncuCyte’s Scratch Wound Cell Migration Module, which normalizes for cell number and calculates relative wound density.

To quantify cell invasion, cells first underwent a double siRNA transfection course at days 0 and 4 of the experiment, as described for the screen. At day 5, cells were trypsinized and seeded at 50,000 cells per insert in 6-well plate invasion chambers coated with Matrigel (354481; Corning Biocoat), covered with serum-free medium, and placed in wells containing media with 10% fetal bovine serum in triplicates. After 30–36 h incubation, uninvaded and invaded cells were detached from both sides of the inserts with trypsin treatment and physical scraping, quantified by Cell Counter (Bio-Rad), and ratios of invaded versus uninvaded cells for each condition were calculated, normalizing for cell proliferation.

Flow cytometry
PANC-1 cells underwent an 8-d course of siRNA knockdown as described for the screen, and at day 8 cells were trypsinized, washed in PBS/2% serum, stained with mouse CD24-PE (555428; BD) or rat CD44-APC (17–0441–82; eBioscience) per manufacturer’s protocol and processed for routine flow cytometry in a FACSCaliber Analyzer (BD).

Tumorsphere formation
PANC-1 cells underwent siRNA transfections as described for the screen at days 1 and 4 of the assay and were subsequently detached at day 5 with trypsin, plated in ultralow attachment 6-well plates (Costar 3471) at a density of 4,000 cells per well, and cultured in stem cell media composed of DMEM supplemented with 1× B27 Supplement, 20 ng/mL human basic fibroblast growth factor, 20 ng/mL epidermal growth factor (all three from Invitrogen), and 4 μg/mL heparin calcium salt (Fisher). Spheres were imaged at day 12 of the assay by using microscopy, and only cell clusters >10 μm were considered spheres.

Cell viability
After double siRNA knockdown at day at days 0 and 4 of the experiment, the following compounds were added at day 5: gemcitabine hydrochloride (3259; Tocris Bioscience) or 5-Fluorouracil (03738; Sigma-Aldrich). Cells were exposed 72 h later to CellTiter-Glo (G7570; Promega) and assessed with a 2104 EnVision reader (PerkinElmer).

Analysis and calculation of IC50 (the half maximal inhibitory concentration) was performed in Prism (GraphPad) by using nonlinear curve fitting regression algorithms.

Knockdown experiments
All knockdown experiments were performed with siRNA sequences from siGENOME (Dharmacon) with the same reagents described for
of histones H3 and H4 in 78 combinations were extracted via Skyline
experiment. Quantitative data on 40 distinct posttranslational modifications
(Q-Exactive HF; Thermo Fisher) in a parallel-reaction monitoring exper-
iment described (Maile et al., 2015). Histone peptides were quantified by
derivatization of peptide N termini with phenyl isocyanate as previ-
spectrometry by propionlylation of lysines, digestion with trypsin, and
derivatization of peptide N termini with phenyl isocyanate as previ-
sely described (Maile et al., 2015). Histone peptides were quantified by
capillary reverse-phase liquid chromatography nanoelectrospray tandem
mass spectrometry on a hybrid quadrupole-orbitrap mass spectrometer
(Q-Exactive HF; Thermo Fisher) in a parallel-reaction monitoring exper-
miment. Quantitative data on 40 distinct posttranslational modifications
of histones H3 and H4 in 78 combinations were extracted via Skyline
software and normalized via internal standards as previously described
(Vinogradova et al., 2016). Modifications on H4K20 are reported here.

ChIP-qPCR

ChIP assays for histone modifications and data analysis were performed as
previously described (Milne et al., 2009) by using EZ Chip (17–371;
Millipore) and an antibody for H4K20me3 (39671; Active Motif). ChIP
signal enrichment was quantified by qPCR by using the following Taqman
assays: FOXA1 Hs01351917_cn, Hs00666612_cn, Hs01162626_ cn;
OVOL1 Hs05268126_cn, Hs02113848_cn, Hs01458465_cn;
OVOL2 Hs02395224_cn, Hs07203924_cn, Hs00475512_cn;
CDH1 Hs05413045_cn, Hs05411756_cn, Hs05397765_cn; EPC
AM Hs04617695_cn, Hs04630814_cn, Hs02913990_cn; KLF4
Hs02900640_cn, Hs02807427_cn, Hs01469971_cn; ZEB1 Hs07440276_ cn,
Hs07442838_cn; TFFC Hs02677106_cn.

Gain-of-function experiments

Coding sequence of full-length wild-type SUV420H2 (National Center
for Biotechnology Information NM_032701.3) and a methyltransferase
deaf allele omitting the sequence 5′-TTCATCAACCATGAC
TGCAACCCACTGCAATCTGGCCCT-3′ from within the SET
domain-coding region were synthesized by GeneArt (Thermo Fisher)
with flanking XhoI and a EcoRI restriction sites and subcloned into
the overexpression vector pIRE2-ZsGreen1 (632478; Clontech). Cells
were transfected with FugeneHD (Promega) and after 72-h BD fluores-
cence-activated cell sorted for GFP-expressing cells.

PDAC immunofluorescence imaging and analysis

Paraffin-embedded sections of human pancreatic cancer were obtained from
Genentech’s tissue bank, deparaffinized, and stained with the nu-
clear stain and antibodies used in the screen, rabbit ZEB1 (sc-25388;
Santa Cruz) with goat anti–rabbit Alexa Fluor 546 secondary anti-
body, and rabbit SUV420H2 (HPA052294; Sigma-Aldrich) with goat
anti–rabbit Alexa Fluor 488 secondary antibody. The latter was also
used on PANC-1 cells grown on glass-bottom chamber slide systems
(154534PK; Thermo Fisher) after fixing in 4% paraformaldehyde. Im-
ages were captured with a confocal microscope (LSM780; Zeiss) by
using Diode 405, Argon 458/488/514, HeNe543, and HeNe633 lasers,
and 5× EC Plan-Neofluar dry 0.16 NA or 20x Plan Apochromat dry
0.8 NA objectives at room temperature with cells or tissues mounted in
Fluoromount-G (Thermo Fisher). Images were analyzed with Zen2010
software (Zeiss), and the fluorescent signal was quantified with ImageJ.

TCGA data

TCGA RNAseq data were downloaded from the Cancer Genomics Hub
at University of California, Santa Cruz, and analyzed with HTSeq
(Andreatta and Reeder, 2014). The results shown here are in part based
on data generated by the TCGA research network. Expression values
were calculated as normalized reads per kilobase gene model per mil-
tion total reads (RPKM), which represent RPKMs values normalized
by an adjusted total library size instead of a simple total read count.
This way of normalization is akin to size-factor normalization, but the
values are reported in the traditional RPKM scale. More details about
this normalization can be found in the study by Srinivasan et al. (2016).

Genentech data

RNAseq reads were mapped to the GRCh 38 genome by using GSNAP
with RefSeq gene models. We used voom+limma (Law et al., 2014) as
implemented in the limma package for the R programming language
and described in the limma user guide to calculate differential gene ex-
pression for cancer versus normal. P-values were corrected for multiple
testing by using the Benjamini-Hochberg method.

Statistics

Analysis and graph preparation was done in Prism 6 (GraphPad). The
statistical method used is indicated in the figure legends for each figure.
For RNAseq analysis, see RNAseq data processing and analysis.

Online supplemental material

Fig. S1 shows images from primary screen described in Fig. 1 for the
siRNA knockdown of various epigenetic targets. Fig. S2 depicts ef-
ciency of SUV420H2 knockdown, the recovery experiment, and effects
of SUV420H2 knockdown on different cancer cell lines of mesenchymal
identity. Fig. S3 depicts representative images of migration assay, the
effects of SUV420H2 knockdown on cell proliferation, FOXA1 protein
levels in pancreatic cancer cells after SUV420H2 knockdown, ZEB1
expression in double-knockdown experiments, ChIP-qPCR experi-
mments for H4K20me3 mark on various genes, and SUV420H2 gain-of-
function experiments on Panc 04.03 cells. Fig. S4 depicts SUV420H2 and
E-CAD immunofluorescent stains in progressive stages of PDAC.
Fig. S5 depicts ZEB1 immunofluorescent stains in progressive stages
of PDAC, transcript levels of SUV420H2 in various cancer types, and
a working model cartoon of SUV420H2 mechanistic control of epithe-
lial/mesenchymal cell states in pancreatic cancer. Table S1 lists siRNAs
used in the study. Table S2 shows RNAseq results for highly and signif-
ically up-regulated genes on SUV420H2 knockdown in PANC-1 cells,
listed by protein coding genes (tab one) and other genes (tab two); tab
three shows full results of RNAseq. Tables S3 and S4 show gene set en-
richment analysis results with MsigDB Hallmark gene sets and with the
MsigDB C2 curated signature collection respectively. Table S5 lists
clinical information for PDAC samples used in this study.

Acknowledgments

All authors were employed by Genentech, Inc., a member of the Roche
group, during their respective contributions to the manuscript.
The authors declare no further conflict of interest.

Author contributions: M. Viotti and I. Mellman designed experi-
ments and wrote the manuscript with input from all authors. M. Viotti,
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