

Comprehensive Analysis of Long Non-Coding RNAs in Ovarian Cancer Reveals Global Patterns and Targeted DNA Amplification

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Abstract

Long non-coding RNAs (IncRNAs) are emerging as potent regulators of cell physiology, and recent studies highlight their role in tumor development. However, while established protein-coding oncogenes and tumor suppressors often display striking patterns of focal DNA copy-number alteration in tumors, similar evidence is largely lacking for IncRNAs. Here, we report on a genomic analysis of GENCODE IncRNAs in high-grade serous ovarian adenocarcinoma, based on The Cancer Genome Atlas (TCGA) molecular profiles. Using genomic copy-number data and deep coverage transcriptome sequencing, we derived dual copy-number and expression data for 10,419 IncRNAs across 407 primary tumors. We describe global correlations between IncRNA copy-number and expression, and associate established expression subtypes with distinct IncRNA signatures. By examining regions of focal copy-number change that lack protein-coding targets, we identified an intergenic IncRNA on chromosome 1, *OVAL*, that shows narrow focal genomic amplification in a subset of tumors. While weakly expressed in most tumors, focal amplification coincided with strong *OVAL* transcriptional activation. Screening of 16 other cancer types revealed similar patterns in serous endometrial carcinomas. This shows that intergenic IncRNAs can be specifically targeted by somatic copy-number amplification, suggestive of functional involvement in tumor initiation or progression. Our analysis provides testable hypotheses and paves the way for further study of IncRNAs based on TCGA and other large-scale cancer genomics datasets.

Citation: Akrami R, Jacobsen A, Hoell J, Schultz N, Sander C, et al. (2013) Comprehensive Analysis of Long Non-Coding RNAs in Ovarian Cancer Reveals Global Patterns and Targeted DNA Amplification. PLoS ONE 8(11): e80306. doi:10.1371/journal.pone.0080306

Editor: Aedín C. Culhane, Harvard School of Public Health, United States of America

Received June 13, 2013; Accepted October 1, 2013; Published November 12, 2013

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Funding: This work was supported by grants from the Swedish Medical Research Council; the Swedish Cancer Society; The Swedish Foundation for Strategic Research; the Assar Gabrielsson Foundation; the Magnus Bergvall Foundation; the Åke Wiberg foundation; and the Lars Hierta Memorial Foundation. Funding for A.J., N.S., and C.S. was provided by the US National Cancer Institute as part of the TCGA Genome Data Analysis Center grant (NCI-U24CA143840 and NCI-R21CA135870) and by a Stand Up To Cancer Dream Team Translational Research Grant, a Program of the Entertainment Industry Foundation (SU2C-AACR-DT0209). The computations were in part performed on high-performance computing resources provided by Swedish National Infrastructure for Computing (SNIC) through Uppsala Multidisciplinary Center for Advanced Computational Science (UPPMAX) under project b2012108. The results published here are in whole or part based upon data generated by The Cancer Genome Atlas pilot project established by the National Cancer Institute and National Human Genome Research Institute (NHGRI). Information about TCGA and the investigators and institutions who constitute the TCGA research network can be found at "http://cancergenome.nih.gov". The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Competing interests: The authors have declared that no competing interests exist.

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Introduction

Recent transcriptomic studies in mammals have revealed an abundance of long non-coding RNAs (IncRNAs) that lie interspersed with coding genes in complex ways [1-3]. LncRNA transcripts typically have mRNA-like properties, such as multiexonic gene structures and poly(A) tails, but lack apparent protein-coding capacity. Although early functional examples (e.g. *H19* [4] and *XIST* [5]) were first described more than 20 years ago, IncRNAs are now emerging as widespread

regulators of cell physiology with diverse roles both in the nucleus and the cytoplasm, including recruitment of histone-modifying complexes to chromatin, regulation of transcription and splicing, and control of mRNA translation.

Several recent studies suggest that IncRNAs may have important roles in oncogenesis [6,7]. For example, *HOTAIR* expression is high in metastatic breast cancer tumors, and its inhibition blocks metastasis in rodent models [8], *MALAT1* expression correlates with metastases and survival in lung cancer [9], and polyA+ transcriptome sequencing (RNA-seq)

recently identified PCAT-1 as a growth-promoting IncRNA in prostate cancer [10]. However, it is notable that the genetic data provided thus far primarily relates to alterations in gene expression. Malignant transformation requires genetic activation of growth-promoting oncogenes and inactivation of tumor suppressors, and this is facilitated in tumors by genomic instability, acquired genetic variability, and clonal expansion [11]. In large cancer genomics datasets, such as those produced by The Cancer Genome Atlas (TCGA) consortium, important cancer genes therefore reveal themselves through striking patterns of recurrent DNA-level alteration, including focal copy-number amplification and deletion [12,13]. However, while IncRNAs and coding genes should in principle be susceptible to activation or deactivation through similar mechanisms, there is thus far little evidence that IncRNAs are specifically targeted by copy-number alterations in cancer independently of proximal coding genes (recently reviewed in

We here performed a large-scale genomic analysis of IncRNAs in high-grade serous ovarian carcinoma (HGS-OvCa). one of the leading causes of cancer death among women in the United States [15], based on high-throughput molecular profiles generated within TCGA [12]. We based our analyses on the comprehensive GENCODE IncRNA catalog, which has been subject to extensive characterization and manual curation [16,17], while using an annotation-unbiased approach where appropriate. We consequently focus on IncRNAs with reproducible expression in independent datasets, based on the assumption that cancer-relevant IncRNAs should, similar to coding genes, have important functions also in normal cells. Using deep coverage RNA-seq data and high-resolution DNA copy-number arrays, we derived simultaneous copy-number profiles and expression data for >10,000 GENCODE IncRNA genes across 407 primary tumors (data available at www.larssonlab.org/tcga-lncrnas). We investigate the global relationship between DNA copy-number and IncRNA expression, and evaluate IncRNAs in relation to established expression subtypes in HGS-OvCa. Moreover, we address whether IncRNAs may be specifically targeted by focal copynumber alteration in cancer.

Results and Discussion

Molecular profiling of IncRNAs in 407 tumors

We used the GENCODE [17] annotation as our main framework to investigate patterns of IncRNA copy-number alteration and expression across 407 stage-II-IV HGS-OvCa tumors [12]. We found that the GENCODE IncRNA subset [16], which encompasses 10,419 manually annotated IncRNA genes (version 11, Figure 1A), showed a high degree of polyadenylation as determined by normal tissue RNA sequencing (RNA-seq) data (Figure 1B). Copy-number data from comparative genomic hybridization (CGH) arrays was available for 486 primary tumors, and the vast majority of GENCODE IncRNAs (97%) expectedly resided in covered regions. We next processed a total of 25.7 billion GENCODEmapped read pairs from polyA+ RNA-seq (on average 63.1 million per sample) to derive expression profiles for all

GENCODE IncRNAs in 407 of these tumors (Figure **1C**). Only 225 million read pairs mapped to IncRNA loci (on average 553,000 per sample), emphasizing the need for high sequence coverage to accurately quantify IncRNAs.

Comparison of length-normalized (RPKM-type [18]) expression values between coding and lncRNAs genes confirmed that lncRNAs were expressed at substantially lower levels (Figure 1D), in agreement with previous reports from a wide variety of cellular sources [1,16,19]. While 87% of coding genes showed an RPKM level >1 in at least one tumor, only 36% of lncRNAs reached this level of expression. Genomewide correlations between DNA copy-number amplitude and RNA levels were lower for lncRNAs compared to coding genes, but this discrepancy was reduced when only considering abundant and frequently copy-number altered genes (Figure 1E). A complementary analysis based on Affymetrix Exon 1.0ST arrays, which can interrogate a subset of lncRNAs, yielded similar results (481 samples, Figure S1 in File S1).

LncRNAs associate with expression subtypes

Previous analysis of coding gene expression in HGS-OvCa identified four robust subtypes, which were termed 'immunoreactive', 'differentiated', 'proliferative' and 'mesenchymal' based on their gene content [12,20]. The subtypes were further shown to be associated with specific genomic alterations, where the proliferative group has a lower frequency of *MYC* amplification and *RB1* deletion, while the immunoreactive group has a higher frequency of *MECOM* amplification. We here tested if established subtypes in HGS-OvCa also have distinct patterns of IncRNA expression.

Tumors with available subtype, clinical and expression data were randomly partitioned into two sets (n=200 each), withholding one half for later validation. We identified 455 lncRNAs that were induced or repressed specifically in one of the four subtypes relative to remaining samples (Figure **2A**, as detailed in Methods). These expression patterns were clearly maintained in the validation set, confirming that subtype associations were non-random (Figure **2B**). In addition, a tumor's expression subtype could be predicted based on subtype-associated lncRNAs using a simple classifier (**Methods**) in the majority (77%) of tumors (Figure **2B**).

Antisense-overlapping IncRNA may show strong positive correlations with their coding hosts [16], motivating analysis based on intergenic IncRNAs alone. We therefore investigated a subset of 152 IncRNAs with intergenic localization, and found their expression patterns to be similar to the full set in the validation data (Figure 2B, Table S1 in File S1, 73% accuracy). Although coding neighbor genes of the 152 IncRNAs were still moderately predictive of subtype (51%), they showed considerably weaker patterns of subtype-specific expression (Figure 2B). Notably, among intergenic IncRNAs induced in the mesenchymal subtype were MIAT/Gomafu, a known target and co-activator of Oct4 with a role in stem cell pluripotency [21]. NEAT1 and UCA1 were repressed in the proliferative subtype: NEAT1 is essential for the structure of nuclear paraspeckles [22] and has been shown to be upregulated in ovarian cancer [23], while UCA1 is a known regulator of cell growth in bladder carcinoma [24]. We additionally assessed IncRNA levels in

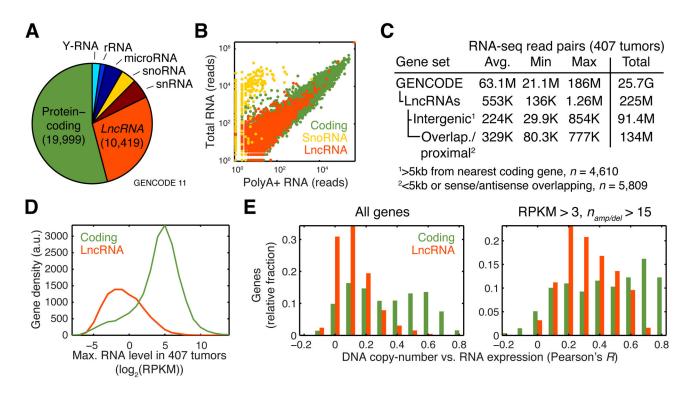


Figure 1. Molecular profiling of IncRNAs in 407 ovarian adenocarcinomas. A, Relative abundances of gene categories in the GENCODE 11 annotation (unique loci). **B**, Polyadenylation status of IncRNAs, determined by polyA+ vs. total RNA-seq from a mixture of 16 tissues. **C**, LncRNA expression profiling using polyA+ RNA-seq across 407 tumors. In total 25.7 billion uniquely mapped read pairs, encompassing >3 terabases, were counted in GENCODE genes. The table shows per-tumor sequencing depth, based on all GENCODE-mapped reads or IncRNA subsets. **D**, Distributions of IncRNA and coding gene expression levels (maximum RPKM in all tumors). RPKM, reads per kilobase per million reads. **E**, Histograms of correlations between DNA copynumber and RNA level (based on 407 tumors with dual data). Left panel: IncRNAs overall (left panel, n = 10,066), showing lower correlations compared to coding genes. Right panel: improved correlations when considering genes expressed at RPKM > 3 (top 19% IncRNAs, n = 1920) that were also amplified or deleted in >15 samples (right panel, n = 125).

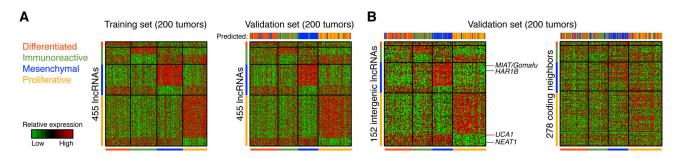


Figure 2. LncRNAs associate with expression subtypes. A, 455 lncRNAs showed increased or reduced expression in one of four previously defined expression subtypes (200 random tumors, left). These lncRNAs maintained their subtype-selective expression patterns in 200 independent tumors, and subtype could there be predicted based on their expression at 77% accuracy (right). **B**, The same analysis based on an intergenic lncRNA subset (*n* = 152, 73% accuracy, left). Nearest up- and downstream protein-coding neighbors of these lncRNAs (278 unique genes) lacked strong subtype-specific expression patterns and their combined signature was less informative of subtype (51% accuracy, right). doi: 10.1371/journal.pone.0080306.g002

relation to patient survival, but did not find reproducible associations. Our results show that expression subtypes in HGS-OvCa, originally defined based on coding gene expression, are each associated with distinct IncRNA expression signatures, and we speculate that these IncRNAs could contribute to transcriptional reprogramming or otherwise act in the cellular circuits that are altered in these tumor types.

LncRNAs in regions of focal copy-number alteration

Tumor genomes are mosaics of chromosomal aberrations, of which some are under selection to activate or inactivate specific oncogenes or tumor suppressors. In large patient cohorts, individual targeted genes may therefore be deduced through patterns of recurrence, in particular when altered regions are narrow (focal) [25]. The GISTIC algorithm, when applied to copy-number data from TCGA ovarian cancer, identified several regions of focal recurrent copy-number alteration [12]. Many of these encompass well-known cancer genes, but in some cases the targets remain poorly defined. We hypothesized that IncRNAs could be drivers in some of these events, and therefore screened narrow focal regions for overlaps with IncRNAs.

There were 35 narrow (overlapping with at most 5 coding genes) amplifications or deletions that were significant at a false discovery rate (residual q) of 0.05 (Figure 3A, Table S2 in File S1). Many overlapped with established proto-oncogenes such as CCNE1 and MYC, and the RB1, NF1 and PTEN tumor suppressors, but IncRNAs were also present together with coding genes in several cases (Figure 3A). Although selection for copy-number alteration at these loci could in principle be explained by IncRNAs, either alone or in combination with their coding neighbors [26], we focused instead on two focal peaks that lacked protein-coding genes: an amplification on 1q25 and a deletion on chromosome 4q34 (indicated with * in Figure 3A). The deleted region was in a large intergenic space ~1 Mb from ODZ3; a gene recently found to be targeted by L1 retrotransposition in colorectal cancer [27] and deleted in neuroblastoma [28]. While deleted segments in HGS-OvCa were clearly separated from ODZ3 and encompassed two annotated IncRNA genes (Figure S2A in File S1), these lacked relevant expression (<5 mapped reads in >99% of tumors), and we failed to reveal other candidates by investigating RNA-seq read coverage in the region (data not shown). Further analysis suggested that the deletions might be indirectly targeted at ODZ3 through disruption of associated regulatory DNA (Figure S2B in File S1), and the region was not further characterized.

Focal somatic amplification of OVAL IncRNA

We next investigated the 1q25 amplification, which was focally gained in 16/407 patients (3.9%), and centered on a 128 kb intergenic region between the *ACBD6* and *XPR1* genes (henceforth the AXI region). The AXI region lacks proteincoding genes, but contains a single annotated lncRNA gene, *RP11-522D2.1*, near its center (Figure **3B**). This lncRNA, which we here term *OVAL* (ovarian adenocarcinoma amplified lncRNA), coincides closely with the focal peak identified by GISTIC, and is positioned 55 kb and 65 kb from the nearest coding protein-coding neighbors. Notably, the amplified

chromosomal segments were often small (50-100 kb) and encompassed the full *OVAL* gene, while being constrained to the AXI region or extending only partially into the neighboring genes (Figure **3B**).

Since the focal DNA amplification pattern pointed to OVAL as the alteration-driving gene in this region, we next investigated if this was supported by the expression pattern of OVAL in the tumors. OVAL expression was low or absent in both normal fallopian tube (Figure S3 in File S1) and in the majority of tumors, including most cases with wide 1q amplification. However, focal amplification of the OVAL locus coincided strikingly with OVAL transcriptional activation (Figure 3C). OVAL RNA was on average 46-fold higher in focal cases compared to remaining samples (P = 3.5e-8, Wilcoxon rank sum test), and OVAL ranked 74th of all GENCODE IncRNAs based on maximum expression in all tumors (Table S3 in File S1). Similar results were obtained using hybridization-based Exon 1.0ST data (Figure S4 in File S1).

Although AXI region focal amplification did not appear to directly target the flanking coding genes, these could still be indirectly affected at the level of gene expression. This would be compatible with their regulatory sequences being altered, or *OVAL* having a *cis*-regulatory role in controlling their transcription. However, neither *ACBD6* nor *XPR1* were notably induced in focally amplified cases (Figure 3D). In addition, these genes are not previously described as altered in cancer, further supporting that *OVAL* is independently targeted in the AXI intergenic region.

Investigation of RNA-seq read coverage in the AXI region revealed that OVAL was the main expressed locus in focally amplified tumors, while remaining samples showed low transcriptional activity in this region (Figure 3E). Although additional transcription was observed outside of the OVAL locus, most notably in the upstream region (Figure 3E), these signals were not consistent between individual tumors (Figure S5 in File S1). Examination of available data in Genbank revealed only a few singular ESTs sequences in the AXI region away from the focal center, while RP11-522D2.1/OVAL was supported by 12 spliced ESTs and 6 cDNA sequences. A putative Y-RNA (predicted from RFAM families), close to the AXI region edge 50 kb from the focal peak, was not supported by cDNA/EST evidence or RNA-seg in tumors or normal tissue (data not shown). We conclude that both HGS-OvCa tumor expression profiles and available cDNA/EST evidence point to OVAL as the main stably transcribed unit in the amplified AXI region.

Gene set enrichment analysis (GSEA) revealed that previously defined targets of P53 (TANG_SENESCENCE_TP53_TARGETS_DN) were significantly elevated in *OVAL* amplified tumors (Figure **3F**). While this indicates that *OVAL* activation may coincide with altered P53 activity, *TP53* mutation status and mRNA levels were similar in both groups. Genes encoding muscle-related contractile proteins (STRUCTURAL_CONSTITUENT_OF_MUSCLE) were enriched among those repressed in *OVAL* amplified tumors.

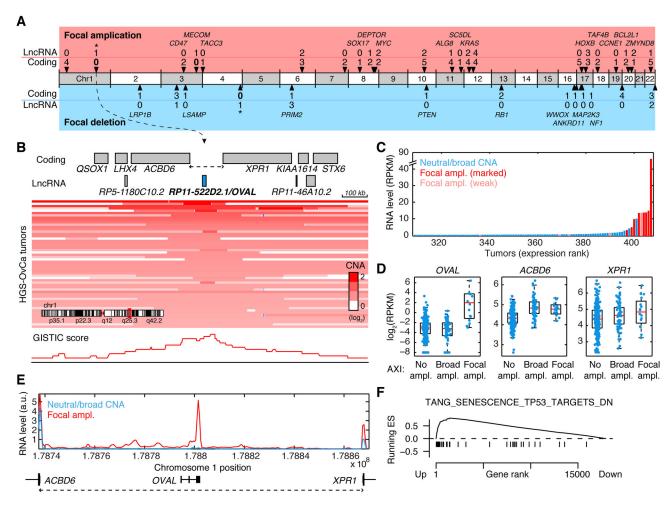


Figure 3. Recurrent IncRNA amplification in serous ovarian carcinoma. A, LncRNAs and coding genes in narrow regions of recurrent amplification or deletion identified by GISTIC (q < 0.05) in 486 HGS-OvCa tumors. Gene counts for 35 tight focal peaks with a maximum of 5 overlapping coding genes are shown. Known cancer genes and unambiguous coding targets are indicated. *, regions investigated in more detail. **B**, Detailed view of the ACBD6-XPR1 intergenic region (AXI region, dotted line) in a ~1 Mb genomic context. The AXI focal peak is centered on RP11-522D2.1/OVAL, an uncharacterized lncRNA on chr1q25. Red shading shows copy-number profiles for individual tumors. **C**, Tumors ordered by OVAL expression. OVAL RNA (y-axis) was mostly low or undetectable, but was induced in focally amplified cases (as defined in Methods). **D**, Neither ACBD6 nor XPR1 were notably induced by AXI region focal amplification. OVAL RNA was low also in broadly amplified cases. ND, not detected. **E**, Average RNA-seq read density in the AXI region (dotted line) for tumors with marked AXI focal amplification (n = 10) compared to remaining tumors (normalized read counts per 1000 nt segment). **F**, GSEA analysis showed than experimentally determined P53 regulated genes are induced in OVAL focally amplified tumors.

Molecular characterization of OVAL

Having established *OVAL* as a likely target in the AXI region, we further characterized this gene in terms of gene structure and normal tissue expression. The *OVAL* gene contains three annotated exons that give rise to a predicted 1489 nt noncoding RNA, where the large third exon contributes most of the sequence. This structure was supported by multiple GenBank mRNA sequences (Figure 4A) and spliced ESTs, as well as RNA-seq from cell lines such as Gm12878 (data not shown). Many of the *OVAL* ESTs and mRNAs originated from human

melanoma cells, and the transcript was consequently pinpointed in a recent bioinformatics screen for melanoma-specific public ESTs [29]. OVAL was also mapped in a recent survey of human intergenic IncRNAs [19], and similar to our own analysis of normal tissue RNA-seq data, this study identified an alternative first exon isoform (Figure 4A) not supported by the tumor expression data. Although additional possible splicing patterns were observed in the tumors, these showed weak and inconsistent expression across samples (data not shown).

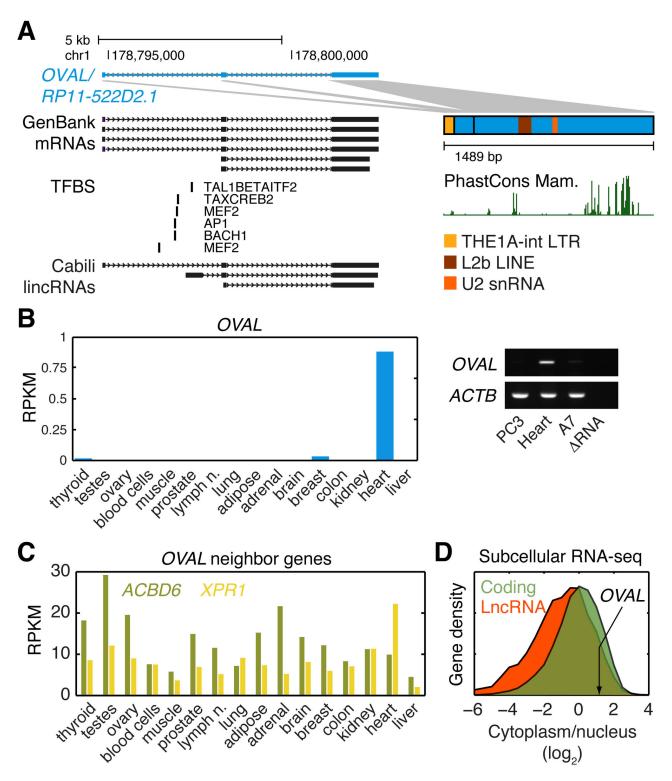


Figure 4. Properties of OVAL IncRNA. A, OVAL locus on chromosome 1. GenBank mRNAs, conserved transcription factor binding sites (TFBS) predicted by the UCSC brower, and other features are indicated. **B**, Normal tissue expression profile of OVAL (RNA-seq). Heart expression was confirmed by reverse transcription PCR. PC3, human prostate cancer cell line; Heart, heart auricle; A7, human melanoma cell line **C**, OVAL and its coding neighbors have disparate expression profiles. **D**, Subcellular RNA-seq from 7 pooled cell lines (Gm12878, HelaS3, HepG2, Huvec, H1hesc, Nhek and K562) shows predominant cytoplasmic localization.

doi: 10.1371/journal.pone.0080306.g004

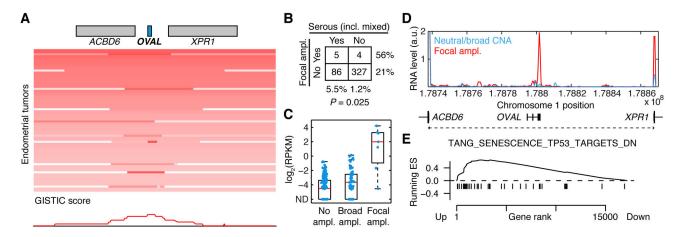


Figure 5. The *OVAL* locus is focally amplified in serous endometrial tumors. **A**, Low frequency focal amplification of the *OVAL* locus in endometrial cancer, but not 16 other TCGA cancers (see Figure **S6** in File S1). **B**, 56% of focally amplified cases were of the serous subtype, compared to 21% overall (P = 0.025, Fisher's exact test). **C**, *OVAL* RNA was strongly induced in a subset of tumors, and this coincided with focal amplification of the AXI region. ND, not detected. **D**, Average RNA-seq read density in the AXI region for tumors with marked focal amplification (n = 4) compared to remaining tumors (normalized read counts per 1000 nt segment). **E**, Similar to ovarian cancer, GSEA analysis revealed induction of P53 targets in *OVAL* amplified tumors.

Evolutionary conservation, based on a mammalian genomic multiple alignment, was low overall, but specific regions in the last exon demonstrated elevated conservation (Figure **4A**). The miRcode database of putative microRNA target sites in lncRNAs [30] revealed a conserved miR-30 site, present in most primates and mammals, in one of these patches. Although the mature sequence is mainly non-repetitive, RepeatMasker [31] identified THE1A-int LTR and L2b LINE derived elements, as well as a possible U2 snRNA sequence in the last exon (Figure **4A**). However, we found no matches to snRNAs or other known structures in RFAM [32], and *OVAL* is therefore unlikely to function as a precursor for a classical structural RNA.

LncRNAs have previously been defined on the basis of codon substitution frequency scores and the lack of an open reading frame (ORF) larger than 100 amino acids [1]. In addition to being classified as non-coding by the GENCODE pipeline, the mature OVAL sequence was non-coding according to the CPC algorithm [33] and using PhyloCSF [34] based on a mammalian alignment. ORFs in OVAL all lack the Kozak consensus and are no longer than 98 amino acids. A recent joint analysis of tandem mass spectrometry data and GENCODE IncRNA sequences, including RP11522-D2.1/ OVAL, identified only one single IncRNA match when excluding misannotated cases, supporting an overall lack of coding capacity for these transcripts [35]. Importantly, no homology to any known protein sequence was revealed by BLASTx analysis. A coding function thus appears improbable, and we conclude that OVAL likely represents a bona fide IncRNA.

RNA-seq from normal human tissues showed that *OVAL* is selectively expressed in heart muscle, and this was confirmed by reverse transcription PCR (Figure **4B**). Two putative conserved myocyte enhancer factor-2 (MEF2) binding sites,

positioned close to the alternative first exon (Figure **4A**), may drive expression in muscle tissue, as both heart muscle and human skeletal muscle myoblast (HSMM) cells exclusively express this alternative isoform (data not shown). The *OVAL* expression pattern is markedly different from its coding neighbors (Figure **4C**), and its subcellular localization was predominantly cytoplasmic (Figure **4D**). This further speaks against a *cis*-regulatory role on nearby genes, and is consistent with our finding that *OVAL* amplification did not notably influence their expression (Figure **4D**). In summary, *OVAL* appears to have a cytoplasmic non-coding function that is independent of its protein-coding neighbors.

OVAL amplification in serous endometrial carcinomas

We next investigated whether *OVAL* amplification is unique to ovarian cancer, and considered copy-number profiles from 16 additional TCGA cancers, ranging in size from 57 to 825 tumors. Interestingly, we observed low-frequency focal amplification of the *OVAL* locus also in uterine corpus endometroid carcinoma, while no obvious focal signal was seen in the remaining cancers (Figure S6 in File S1). Closer inspection revealed that the focal peak again coincided closely with the *OVAL* gene (Figure **5A**).

A fraction of endometrial tumors are classified as serous or serous-like. These have a close morphological resemblance to their ovarian counterpart [36], and are also genetically similar to ovarian cancer [37]. Consequently, we observed that tumors of the serous subtype were >4 times as likely to carry OVAL focal amplification compared to non-serous tumors (5/91 vs. 4/331, P=0.025, Figure **5B**). Similar to ovarian cancer, focal amplification of the AXI region was associated with strongly increased expression of OVAL (Figure **5C**), and RNA-seq read coverage showed that OVAL was the main transcribed unit in

the region (Figure **5D**). GSEA analysis revealed that experimentally determined P53 regulated genes were upregulated in *OVAL* amplified samples, replicating our previous results from ovarian cancer (Figure **5E**). Taken together, results from ovarian and endometrial cancer suggest that *OVAL* amplification is selected for specifically in serous tumors irrespective of tumor site.

Conclusions

LncRNAs have previously been assayed in clinical materials using next generation sequencing, including a recent study of 64 carcinomas and sarcomas using 3' end sequencing [38], and transcriptome sequencing of 102 prostate tissues and cell lines [10]. In addition, IncRNAs were profiled in normal and cancer tissues based on 272 public SAGE libraries [39]. The present analysis is the first to make use of TCGA RNA-seq to profile IncRNAs in cancer, and to facilitate future investigation we make IncRNA molecular profiles for TCGA tumors available at www.larssonlab.org/tcga-Incrnas.

There is only limited evidence for somatic focal copy-number alteration of IncRNAs in cancer, and described cases involve IncRNAs that are co-altered with proximal coding cancer genes. Two IncRNAs in the *LSAMP* tumor suppressor locus on chromosome 3q13, *OC285194* and *BC040587*, were frequently focally deleted in osteosarcoma, often together with *LSAMP* [26]. These IncRNAs are coexpressed with *LSAMP*, and the three genes are likely functionally interconnected. The *PVT1* locus on 8q24, which gives rise to a variety of spliced noncoding RNAs, is often co-amplified with the nearby *MYC* oncogene [40,41]. However, in time it has become clear that *PVT1* encodes several microRNAs, and its primary role could therefore be that of a microRNA precursor [42,43].

In the case of RP11/522D2.1/OVAL, several independent observations nominate it as an independent target of somatic gene amplification. It is located at the center of a narrowly amplified intergenic segment that lacks other annotated genes, and the focal peak closely coincides with the OVAL gene. RNA-seg read coverage, as well as available cDNA and EST evidence, failed to reveal other credible candidates in the region. Focal, but not broad, amplification coincided with strong induction of OVAL RNA. OVAL was not co-expressed with its coding neighbors, none of which are previously associated with cancer and the closest being more than 50 kb away, and OVAL amplification did not notably alter their expression. This, in combination with a predominantly cytoplasmic localization, speaks against OVAL having a cis-regulatory role on neighbor genes. The replication of these patterns in serous endometrial cancer reinforces the hypothesis.

Focal amplification of the AXI region is relatively rare (3.9%), and amplitude gains typically low. However, HGS-OvCa is characterized by great mutational diversity, with a relative lack of single frequent driving alterations, and the frequency is comparable to known functional alterations such as BRCA1 and BRCA2 somatic mutation (3.5% and 3.2%, respectively [44]). Even single-copy gains in the AXI region were associated with strong induction of *OVAL* RNA, similar to e.g. *IGF2* in colorectal cancer [13]. This implies that copy-number gain

coincides with non-copy-number mechanisms to activate transcription in these tumors. Several recent studies point to cytoplasmic roles for IncRNAs, including posttranscriptional regulation by complementary base pairing with mRNAs [45,46] and inhibitory binding to microRNAs [47-49]. Future experimental studies are needed to unravel the putative cytoplasmic function of *OVAL* RNA, and to understand how its hyperactivation may contribute to serous tumor development.

LncRNAs have until now received little attention in large-scale cancer genomics efforts such as TCGA. This report complements the coding-centric framework used for the original TCGA HGS-OvCa analysis, and paves the way for futures studies of lncRNAs based on these powerful datasets.

Methods

GENCODE annotation and IncRNA definition

Tab-delimited files and BED files, describing the GENCODE V11 gene annotation, were obtained through the UCSC browser (the most current release available on May 11 2012 for the hg19 assembly). GENCODE 'comprehensive' and 'pseudogene' sets were merged, and transcripts with ambiguous genome mapping were removed (n = 330), resulting in a set of 179,526/53,433 transcripts/genes with unique mapping to a single locus. To define a IncRNA subset, we relied on the coding/non-coding classification provided by the GENCODE/ENSEMBL pipeline, and considered as IncRNAs genes that exclusively produce transcripts of the 'antisense', 'lincRNA', 'non_coding' and 'processed_transcript' types. Genes producing non-coding mature transcripts shorter than 200 nt were excluded, as were genes with symbols matching any coding genes in the RefSeg or UCSC xenoRefGene set. This removed a small number of cases in GENCODE of obvious incorrect coding/non-coding classification, and resulted in a final set of 15,977/10,419 IncRNA transcripts/genes. An 'intergenic' IncRNA subset was further defined by determining, for each gene, the smallest distance from either the 5'-most transcript start or the 3'-most transcript end to the nearest coding GENCODE gene, and requiring this distance to be >5 kb.

Genomic copy-number data

We used segmented genomic copy-number data for 486 unique patients, produced on the Agilent 1M platform and segmented using binary circular segmentation as described in the original TCGA ovarian study [12]. As these were generated using the Hg18 genome assembly as reference, we used the LiftOver tool (http://genome.ucsc.edu/util.html) to remap the GENCODE gene annotation from Hg19 to Hg18. Genes were assigned copy-number amplitudes (log₂ scale) by comparing gene coordinates with segment coordinates to identifying overlapping segments. In cases of partial overlap with several segments the minimum amplitude was chosen, the rationale being that amplification of the complete gene would normally be required to increase its activity, while partial deletion would normally disrupt its function. Genes that had undetermined copy-number amplitudes in more than 50% of samples were excluded, resulting in a final copy-number matrix covering

53,433 genes of which 10,066 were lncRNAs, 11,316 were pseudogenes and 19,061 were protein-coding. Segmented copy-number data for endometrial tumors (n=443, Hg19 assembly), produced on the Affymetrix SNP6 platform, were obtained through the TCGA data matrix and processed using the same pipeline.

TCGA RNA-seq processing

RNA-seg sequence libraries in BAM format (2x75 nt pairedend reads) for 412 primary HGS-OvCa tumors were downloaded from cgHub (http://cghub.ucsc.edu, data available on Oct 9 2012). The BAM files were produced by the BCCA Genome Science Center TCGA RNA-seq pipeline, which briefly uses BWA [50] for alignment to the Hg18 genome assembly and to exon junctions derived from Ensembl/ GENCODE, UCSC genes and RefSeq. Low-quality alignments (mapping quality 0) were removed and sequences were namesorted and converted to SAM format using SAMtools [51]. We used TopHat [52] with default parameters to realign a subset of the samples to enable unbiased study of splicing patterns in the AXI region. TCGA endometrial RNA-seq data in BAM format (76 nt single-end reads) for 321 tumors was obtained from cgHub (downloaded on Oct 18 2012). These BAM files are not directly useful for quantifying GENCODE IncRNAs as they were generated by alignment to a limited transcriptome database. They were therefore converted to FASTQ format and realigned to the Hg19 assembly with TopHat using the "-G" option with known splice junctions from GENCODE. Read counts for individual GENCODE genes were subsequently determined using HTSeq-count (http://www-huber.embl.de/ users/anders/HTSeq) in "intersection-strict" mode, by considering only uniquely mapped reads. RPKM expression levels for IncRNAs (n = 10,419) and other GENCODE genes were finally calculated by normalizing for mRNA length and library size as determined by the number of GENCODEmapped reads. For analyses requiring log2-scale values, a pseudo value of 0.01 was added before conversion to avoid log of zero [53]. HGS-OvCa samples with less than 20 million GENCODE-mapped read pairs and without matching copynumber data were excluded, resulting in a final set of 407 tumor expression profiles with on average 63.1 million GENCODE-mapped read pairs each (25.7 billion in total). For endometrial samples, 10 million GENCODE-mapped reads were required, for a resulting final set of 293 tumors with on average 19.1 million GENCODE-mapped reads (5.6 billion in total). Expression coverage plots were generated by dividing the genome into partially overlapping 1000 nt tiles spaced 500 nt apart. Tile read counts were determined using BEDTools (coverageBed utility) [54], and these were normalized based on the median of the top 5% expressed tiles in each sample.

Exon array processing

565 Affymetrix Human Exon 1.0ST array (HuEx) CEL files for primary HGS-OvCa tumors were obtained from the TCGA (level 1 data). Probeset intensities (1,411,399 probesets) were determined using Affymetrix APT software suite using the RMA-sketch algorithm. Genomic probeset locations (Hg19 assembly) for the HuEx array were obtained from Affymetrix,

and subsequently mapped to exons in our GENCODE-derived gene annotation. For each gene, all exonic probeset signals (linear scale) were averaged to produce a single gene expression value. The final expression matrix contained expression levels for 45,953 genes of which 8,777 were lncRNAs, 9,902 were pseudogenes and 19,774 were proteincoding.

Screening for IncRNAs in focal regions

Focally amplified or deleted genomic loci, identified by the GISTIC algorithm [25] as described previously [12], were screened for overlaps with annotated IncRNAs. We considered the "wide peak" as defined by GISTIC, and emphasized those that contained annotated IncRNAs while lacking coding genes. For integrative analyses requiring cases of OVAL/AXI region focal amplification to be defined, we choose the following rules: Tumors were classified as focal if the difference between the AXI center copy-number amplitude (CNA) and the minimum CNA in the upstream ACBC6 region (log₂ scale), as well as the downstream XPR1 region, was >0.2, or alternatively >0.4 (marked focal amplification). A CNA threshold of 0.4 was used to define broadly amplified samples. To investigate putative focal amplification of the OVAL locus in other cancers than ovarian, GISTIC results for additional TCGA cancers were obtained from the Broad Institute Firehose pipeline. We considered 16 additional cancers where copy-number data was available for at least 50 patients (ranging from 57 to 825) at the time of download (Sept 19, 2012). Amplified segments and GISTIC scores were visualized with IGV.

Additional RNA-seq analyses

Raw reads from the Illumina BodyMap2 RNA-seq expression compendium, including data from 16 individual normal human tissues (2x50 nt, unstranded), as well pooled RNA from the same tissues (100 nt, stranded), was obtained from EBI's ArrayExpress repository (accession E-MTAB-513). To determine the polyA status of IncRNAs, sequence traces from DSN-normalized pooled total RNA (148 million reads) and from DSN-normalized polyA+ pooled RNA (287 million reads) were mapped to the Hg19 assembly using TopHat and quantified using HtSeq-count as described above for TCGA datasets, after stripping of adapter sequences using the FastX toolkit (http://hannonlab.cshl.edu/fastx_toolkit). Tissue expression profiles were determined by applying a similar workflow on reads from the 16 tissues (2.54 billion reads in total).

To determine the subcellar localization of *OVAL*, subcellular RNA-seq reads produced within the ENCODE project [55] were obtained through the UCSC FTP server. A total of 1.71 billion 76 nt read pairs from polyA+ nuclear and cytoplasmic fractions from seven human cell lines (Gm12878, HelaS3, HepG2, Huvec, H1hesc, Nhek and K562) were aligned to the human hg19 reference genome with Tophat. Ratios between pooled normalized read counts from nuclear and cytoplasmic fractions were finally calculated for all GENCODE genes.

Subtype and gene set enrichment analyses

We used the Matlab CGDS toolbox [44] to access TCGA subtype data within the Matlab (Mathworks Inc.) environment.

Subtype-specific lncRNAs were identified using the *t*-statistic based on log-scale expression values, by testing for differential expression between tumors in one subtype and remaining tumors. Genes with more than 50% undetermined values (0 reads) were excluded. A *t* threshold of 4/-4 was applied to find subtype-associated lncRNAs, and lncRNAs that were ambiguously assigned to more than one subtype were not considered. Log-scale expression values for each gene were z-score normalized prior to visualization. Prediction of subtypes from lncRNA expression profiles was based on a simple scoring system, where the difference between the means of z-score normalized expression values of previously defined subtype-induced and sub-type repressed lncRNAs was calculated. This was done separately for each subtype, choosing the highest scoring subtype in each tumor.

We used GSEA [56] to identify differentially expressed gene sets in the tumor samples with focal OVAL amplification. We quantified differential expression between two phenotypes (amplification vs. rest) using the t-statistic, and considered all gene sets (MSigDB) of size 15-500 genes (n = 5332 gene sets). Gene set enrichment P-values were computed with respect to a null distribution obtained from 1000 randomizations of the patient-phenotype labels.

Reverse transcription PCR

RNA from heart auricle was a kind gift from Dr. Elin Stenfeldt, and RNA from A7 melanoma cells (ATCC) and PC3

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prostate cancer cells [57] were kindly provided by Dr. Levent Akyurek. Total RNA was reverse-transcribed using the High Capacity cDNA kit (Applied Biosystems). PCR (55 annealing temperature, 35/30 cycles for OVAL/ACTB) was performed using the following primers: RP11-522D2.1/OVAL: 5'-AGGCCAATATGCAGACAAGG-3' and 5'-AGTTCTCCAGTGGGGGTCTT-3'; ACTB: 5'-ACTCTTCCAGCCTTCCTTCC-3' and 5'-GTACTTGCGCTCAGGAGGAG-3'.

Supporting Information

File S1. Supplementary figures and tables. (PDF)

Acknowledgements

We wish to thank the ENCODE consortium and the Gingeras group (Cold Spring Harbor Laboratories) for generating subcellular RNA-seq data used in this study, and Illumina for making their BodyMap2 expression dataset publicly available.

Author Contributions

Conceived and designed the experiments: EL NS CS JH AJ. Performed the experiments: EL. Analyzed the data: EL RA AJ. Wrote the manuscript: EL.

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