Introduction

Progress in treating human prostate cancer has been hampered by the finding that histologically identical cancers exhibit widely variant clinical behavior. It is anticipated that the development of molecular taxonomies of prostate cancer will lead to the definition of unique cancer subsets each marked by different clinical behavior, probabilities of response to therapy, and cancer biology. Genomic efforts to render such molecular classifications have included expression profiling, proteome analysis, and comparative genomic hybridization. These technologies typically require nonfixed cancer tissue and often substantial quantities of cancer tissue. Ultimately, the widespread clinical use of cancer classification methods will require technologies capable of assaying small foci of cancerous tissue isolated from frozen tissues. A novel informatics platform, dChipSNP, was used to automate the definition of statistically valid regions of LOH, assign LOH genotypes to prostate cancer samples, and organize by hierarchical clustering prostate cancers based on the pattern of LOH. This organizational strategy revealed apparently distinct genetic subsets of prostate cancer.

Abstract

Oligonucleotide arrays that detect single nucleotide polymorphisms were used to generate genome-wide loss of heterozygosity (LOH) maps from laser capture microdissected paraffin-embedded samples using as little as 5 ng of DNA. The allele detection rate from such samples was comparable with that obtained with standard amounts of DNA prepared from frozen tissues. A novel informatics platform, dChipSNP, was used to automate the definition of statistically valid regions of LOH, assign LOH genotypes to prostate cancer samples, and organize by hierarchical clustering prostate cancers based on the pattern of LOH. This organizational strategy revealed apparently distinct genetic subsets of prostate cancer.

Materials and Methods

Prostate Cancer Samples and Clinical Data. Paraffin blocks along with frozen SVs were available for 50 of a set of 52 prostate cancer samples, the clinicopathological characteristics of which were described previously (8).

DNA Isolation. Half of each snap-frozen SV sample was fixed in 10% buffered Formalin acetate, mounted in Tissue-Tek OCT, H&E stained, and visualized by light microscopy to confirm the absence of cancer. Germ-line DNA was prepared from the remainder by proteinase K digestion in SDS buffer followed by two extractions in Tris-buffered phenol and chloroform extraction. DNA was precipitated with 1/10th volume of 3.5 M sodium acetate, 20 μg of glycogen, and 2.5 volume of ethanol; washed in 70% ethanol twice; and resuspended in 50 μl of TE.

Cancer DNA was obtained from the corresponding formaldehyde-fixed paraffin-embedded prostatectomy cancer tissue. In brief, cancer epithelial cells were identified by pathologists (M. Loda and M. Lechpammer) and retrieved by LCM using an Arcturus PixCell II system (Mountain View, CA). Before DNA extraction, LCM adhesive caps were visualized by light microscopy to ensure ≥90% purity of epithelial cells. Cancer cells adhered to the LCM cap were incubated in 50 μl of digestion buffer (1% Tween 20, 1 mg/ml Proteinase K, and 1 × TE) overnight at 37°C, heated to 95°C for 10 min, and clarified by centrifugation. DNA was quantitated using the PicoGreen double-stranded DNA Quantitation Kit using 2 μl of DNA (Molecular Probes; P-11495). The sample volume was split in two and subjected to SNP PCR amplification according to the manufacturer’s protocol without further DNA purification.

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LOH are discussed and presented in full in a forthcoming study.\(^6\) Briefly, within a 6 Mb window, a summary LOH score was derived across the entire set of cancer/normal comparisons. The summary LOH score was compared with the same score calculated from randomly permuted data. Regions of LOH differing statistically in the actual data from that derived from the permuted data were identified, and LOH assignments were made for each cancer based on the calls made for specific SNP alleles within the bounded region. Hierarchical clustering based on LOH calls for these specific regions was carried out as described previously \(^9\).

**Results and Discussion**

Fifty-two prostate cancers, obtained at the time of radical prostatectomy, were studied previously by expression profiling to predict outcome after radical prostatectomy and detect the presence of an expression-based metastatic signature \(^8, 10\). For each cancer, paired normal germ-line DNA was purified from pulverized unfixed frozen SV obtained at the time of surgery. Sections of each SV were examined histologically and found to be free of cancer.

The genotypes for 1494 SNPs were obtained using 120 ng of SV as the normal DNA. As described previously, each SNP allele was specifically amplified in 24 pooled highly multiplexed PCR reactions, and after reamplification and labeling, the genotype of each allele was determined by hybridization to the Hu2K SNP array \(^2\) (Affymetrix) using the standard HuSNP protocol \(^3\) (as described in Lindblad-Toh, 2000). The rate of alleles successfully called (call rate) on the array is an indicator of the quality of the DNA extraction, amplification, and hybridization procedures and determined using the MAS software \(^2\) (Affymetrix). The average call rate was 83.2% and ranged from 74.8 to 87.7% for assays run on the SV DNA. These results are consistent with data published previously (Fig. 1).

In parallel, paraffin-embedded cancer blocks were retrieved for each patient, sectioned, and reviewed by experienced prostate cancer pathologists (M. Loda and M. Lechpammer). Pure populations of malignant prostate epithelial cells were obtained using a Pixell LCM instrument. For each cancer, cells were collected onto the LCM cap until the cap was approximately half full. DNA was extracted in SDS-proteinase K buffer and used without further purification. Half of the LCM material, ranging from 4.6 to 47.1 ng of DNA, was then used in the SNP amplification protocol without further adjustment.

For each patient, sectioned, and reviewed by experienced prostate cancer pathologists (M. Loda and M. Lechpammer). Pure populations of malignant prostate epithelial cells were obtained using a Pixell LCM instrument. For each cancer, cells were collected onto the LCM cap until the cap was approximately half full. DNA was extracted in SDS-proteinase K buffer and used without further purification. Half of the LCM material, ranging from 4.6 to 47.1 ng of DNA, was then used in the SNP amplification protocol without further adjustment.

SNP allele call rates in the cancer samples (81.6 ± 3.3%) did not vary significantly from that obtained with the normal DNA (Fig. 1). Thus, the quality of SNP genotyping with LCM-captured DNA was comparable with that obtained with DNA extracted by standard methods (Fig. 1). Using these methods, LOH detection was carried out in 50 available cancers from the original set of 52 cancers studied previously. The raw data are available on the Internet.\(^7\)

To enable the analysis of the SNP array data across multiple cancer–normal comparisons, a new informatics platform known as dChipSNP was developed. This platform is based on dChip, a bioinformatics package for oligonucleotide array-based expression analysis \(^11, 12\). The details of the statistical methods used in the dChipSNP program for LOH determination are described in a forthcoming study.\(^5\) dChipSNP integrates publicly available SNP, gene, and cytoband mapping information with automated methods for detecting statistically meaningful regions of LOH from a series of paired normal and cancer SNP genotypes. dChipSNP is available for download on the Internet.\(^8\)

\(^{6}\) M. Lin et al., submitted for publication.
\(^{7}\) Internet address: http://research.dfcii.harvard.edu/sellerslab/datasets/index.html.
\(^{8}\) Internet address: http://www.dchip.org.
of chromosomal areas where all heterozygous and thereby informative alleles are rendered homozygous in the cancer. The boundaries of such regions of LOH are defined by either the presence of retained heterozygous alleles, the ends of chromosomal arms, or the centromere. This classical definition assumes that all data points are completely accurate and that all polymorphic alleles are mapped correctly within the genome. Array-based methods of SNP detection may have a certain degree of inaccuracy ("noise"), and moreover, the precise genome mapping of each SNP is still not completely stable. Thus, "true" regions of LOH can be interrupted by apparently false positive "retained" SNP alleles. Conversely, true regions of retention of heterozygosity may be interrupted by false LOH calls. To take these concerns into consideration, a method for assessing the probability of LOH within a given set of normal cancer comparisons was used where within a sliding 6 Mb window, a summary LOH score was derived. To determine whether the measured summary LOH scores exceed those that might occur by chance alone, 1000 data sets were generated in which the sample designation of cancer versus normal was randomly permuted. Regions of deletion were identified where the summary LOH score exceeded that found in the permuted data sets. A P-value, corrected for multiple hypothesis testing, of 0.05 was used as the cutoff (as described in detail in Lin et al.). The predicted region of LOH was demarcated by dChipSNP. Fig. 2A shows the application of this platform to this data set, specifically showing a demarcated region of deletion predicted on chromosome 8 (8p21).

Fig. 2. SNP array-based LOH maps for 8p21 and 10q23. A, automated detection and demarcation of the 8p21 region of LOH in prostate cancers using dChipSNP. Each column represents one tumor/normal pair. Along the far right, the position and name of each SNP are shown. Where the density of SNPs is too high to allow a full display of all SNPs, hidden SNPs are indicated by small blue dots. Additionally along the right within the shaded gray box is the LOH score indicated as a blue line and the P = 0.05 threshold indicated as a red line. Tumors in which there is LOH within the 8p21 region are identified and organized by hierarchical clustering (left branch) as distinct from tumors with definitively retained alleles within the 8p21 regions (right branch). There are no uninformative tumors for this region. Finally, clinical or pathological information pertinent to each tumor is indicated along the top (GS, Gleason Score; CStage, Clinical Stage; PStage, Pathological Stage; Cp Inv, Capsule invasion; SV Inv, SV invasion; LN +, lymph node metastases; Marg +, surgical margin positive; Fail PSA, failure after prostatectomy; LV Inv, lymphatic vessel invasion; and risk, prostate cancer risk group). B, automated detection and demarcation of the 10q23 region of LOH. dChipSNP precisely demarcates the area centered over the PTEN tumor suppressor gene. Left panel, a segment of the LOH plot for chromosome 10 (as in Fig. 2A); however, only those tumors with LOH are shown. dChipSNP allows an expanded zoom in view, and this reveals that the segment of chromosome 10 identified precisely overlies the PTEN tumor suppressor gene.
LOH in prostate cancer localized to 1p, 8p21, 10q23, 16q23, and 17p13 have all been reported previously by multiple investigators, with LOH of 8p21 being among the most frequently reported in prostate cancer (Refs. 13–27 and reviewed in Ref. 28). This methodology, given sufficient overlapping alterations, might lead to enhanced localization of putative genes targeted within the regions of LOH. The *PTEN* gene is thought to be the tumor suppressor gene targeted by alterations in the 10q23 cytoband. The density of informative SNPs in this region, while relatively low, was nonetheless sufficient to give a peak prediction of LOH that corresponded directly with the physical location of *PTEN* (Fig. 2B). Together, the detection of known regions of LOH and colocalization of a predicted peak with a known cancer suppressor argue that microarray-based LOH detection and the automated statistical methods developed for analyzing these data are robust.

Next, we sought to determine whether sample organization algorithms, such as hierarchical clustering, could be applied to this data set. Here, the direct use of the actual SNP allele data in clustering algorithms, much as one would use expression array data, would lead to sample organization based on the germ-line genetic similarities of the individuals rather than on the genetic similarities of the cancer. Moreover, LOH of the same chromosomal region can be marked by completely different sets of heterozygous and thus informative SNPs in distinct cancer–normal pairs. Thus, it is necessary to render the SNP data for each cancer–normal pair into an LOH-based data set. To this end, for each cancer–normal pair, a designation of LOH, retained or uninformative, was made based on the actual genotype of the heterozygous SNPs within the seven regions of significant LOH identified above. Here, retention (RET) was defined as the presence of an informative allele within the region of interest in which heterozygosity was retained (RET) without any heterozygous alleles showing LOH. In dChipSNP, these data can be output as specific tab delimited data sets (see Supplemental Table 1). For purposes of display only, the regions of LOH or retention for each cancer–normal pair are projected as a heat map of blue or yellow using a 50% extension to the next nearest heterozygous SNP. In certain instances, the heat map extending from an allele showing LOH extends into the region of LOH; however, if the actual SNP allele is not within the demarcated region, than the cancer is not designated as LOH for that region.

After such designation, one can cluster cancers based only on a single region of LOH in the chromosome view (Fig. 2) or using all regions of loss defined as significant in the whole genome view. In the first instance, hierarchical clustering was driven by using the comparisons LOH-to-LOH, LOH-to-RET, and RET-to-RET. This resulted in either two or three clusters of tumors for each region of LOH, specifically a cluster with LOH, a cluster with retention, and a cluster of uninformative tumors (Fig. 2A and Suppl. Fig. 1, A–F). This analysis allows the automatic assignment of the genotype for this region to each individual tumor and can be then used in further downstream analysis pertinent to the specific region, e.g., we have looked for gene expression signatures associated with specific regions of LOH by using those tumors that are definitely scored as LOH and those that are definitively scored as RET, while ignoring those that are uninformative in a given region.

For many human cancers, it is thought that there is a sequential progression of accumulated genetic events, ultimately cumulating in a metastatic cancer. In prostate cancer, it remains to be seen whether the heterogeneous nature of the disease can be understood based on this model wherein early genetic events would be then accompanied or not accompanied by late genetic alterations, thus leading to differences in the disease, or whether an alternative, but not mutually exclusive, model in which distinct “parallel” sets of genetic alterations might occur, leading to unique genetic cancer subtypes that have distinct clinical behavior.

To determine whether genome-wide LOH mapping might provide an answer to this question, hierarchical clustering was used to look for separation or cosegregation of LOH events. In this analysis, clustering was enacted using the seven regions of LOH passing the *P* = 0.05 threshold. Again, each tumor was assigned a lost or retained designation based on SNP alleles within the boundaries of LOH. The
comparisons of LOH-to-LOH and LOH-to-RET were used, whereas RET-to-RET was ignored. This allows clustering to be driven primarily by the similarity or difference in a deletion rather than the similarity in retention.

A number of cancers had either no informative alleles or did not show LOH in any of the seven regions and thus were not clustered in this analysis (single branches in Fig. 3. The clustering of tumors with sufficient information revealed that samples were clustered into distinct branches and that these branches contained tumors enriched for specific nonoverlapping regions of heterozygosity. These data suggest the possibility that there are distinct genetic subsets of prostate cancer that can be defined based on LOH analysis. Validation of this observation on higher density and thus more sensitive SNP arrays is an important next step.

In summary, we provide a novel and robust tool for LOH detection. This informatics package and the automated detection of LOH will be a vital tool for the analysis of large SNP LOH data sets that will be available on the second generation of SNP arrays where the cancer-normal genotypes of >10,000 SNP alleles will be available.

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References

Genome-wide Loss of Heterozygosity Analysis from Laser Capture Microdissected Prostate Cancer Using Single Nucleotide Polymorphic Allele (SNP) Arrays and a Novel Bioinformatics Platform dChipSNP

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