INTRODUCTION

The eukaryotic nuclear genome is organized into nucleosomes, the fundamental units of chromatin. The positions of nucleosomes on DNA regulate protein-DNA interactions and in turn influence DNA-templated events. Despite the increasing number of genome-wide maps of nucleosome position, how global changes in gene expression relate to changes in nucleosome position is poorly understood. We show that in nucleosome occupancy mapping experiments in maize (Zea mays), particular genomic regions are highly susceptible to variation introduced by differences in the extent to which chromatin is digested with micrococcal nuclease (MNase). We exploited this digestion-linked variation to identify protein footprints that are hypersensitive to MNase digestion, an approach we term differential nuclease sensitivity profiling (DNS-chip). Hypersensitive footprints were enriched at the 5' and 3' ends of genes, associated with gene expression levels, and significantly overlapped with conserved noncoding sequences and the binding sites of the transcription factor KNOTTED1. We also found that the tissue-specific regulation of gene expression was linked to tissue-specific hypersensitive footprints. These results reveal biochemical features of nucleosome organization that correlate with gene expression levels and colocalize with functional DNA elements. This approach to chromatin profiling should be broadly applicable to other species and should shed light on the relationships among chromatin organization, protein-DNA interactions, and genome regulation.
A modification of nucleosome-mapping assays, the digestion of chromatin to different degrees with MNase, has recently been shown to result in variability of nucleosome occupancy profiles in yeast (Weiner et al., 2010; Henikoff et al., 2011; Xi et al., 2011). Specifically, numerous specific regions of the yeast genome contain footprints that show significantly higher signal under light-digest conditions than under heavy-digest conditions, complicating interpretation of genome-wide nucleosome occupancy mapping studies. These differences have been attributed to an altered chromatin conformation that renders these nucleosomes hypersensitive to MNase cleavage (Henikoff et al., 2011). These so-called “fragile nucleosomes” partially overlap with regions enriched in histone H2A.Z-containing nucleosomes (Xi et al., 2011) and are associated with highly transcribed genes in yeast (Weiner et al., 2010). Fragile nucleosomes have been identified at individual loci in other organisms (Lombrana et al., 2013). While thousands of fragile nucleosomes have been identified in yeast, the widespread occurrence and significance of such fragile nucleosomes remain largely uncharacterized in multicellular organisms, particularly plants.

Here, we show that technical variability in preparing nucleosomal DNA, particularly with respect to the degree of digestion with MNase, leads to variability in nucleosome occupancy signals at specific regions in the maize genome. Exploiting the degree of digestion as an experimental variable in nucleosome occupancy mapping, an approach we term differential nuclease sensitivity digestion as an experimental variable in nucleosome occupancy with MNase, leads to variability in nucleosome occupancy signals, particularly with respect to the degree of digestion, which was not associated with a particular buffer and was not consistent between replicates. However, recent studies in yeast have shown that the degree of nuclease digestion used to isolate nucleosomal DNA can have an influence on nucleosome occupancy profiles in particular genomic regions (Weiner et al., 2010; Henikoff et al., 2011; Xi et al., 2011). In order to determine whether the hypervariable regions we observed might have resulted in part from variation in the extent of MNase digestion, we mapped nucleosome occupancy in earshoots using three different concentrations (hereafter referred to as light, moderate, and heavy; Figure 2B) of MNase as the controlled variable. Plotting of nucleosome occupancy profiles as a function of digestion levels (Figure 2C) exhibited many of the same localized regions of signal variability, but this variability typically was associated with the degree of digestion and was consistently observed across the four replicate digestions.

To examine the buffers’ effects on microarray-based nucleosome occupancy profiles, nuclei extracted in the three different buffers were digested with MNase to near completion, in which the majority of DNA was present in the mononucleosomal (100 to 200 bp) fraction. The 100- to 200-bp DNA fragments were gel-extracted, fluorescently labeled with Cy3, and cohybridized with Cy5-labeled genomic DNA to a custom high-density NimbleGen microarray that was designed to cover the transcription start site (TSS) regions of 400 maize genes primarily derived from the list of Classical Maize Genes (Schnable and Freeling, 2011). Nucleosome occupancy profiles were generated by plotting the log2 ratio (nucleosomal/genomic) of hybridization signals from the microarray (Figure 1C). Nucleosome occupancy scores were very similar for representative genes regardless of which buffer was used. This similarity extended across the majority of the data set in pairwise buffer comparisons (r between 0.88 and 0.92; Figure 1D). Nucleosome occupancy profiles were also generated from 9-d-old seedling shoots and were also highly correlated to immature ear nucleosome occupancy profiles (Figure 1E). The uniformity of nucleosome occupancy profiles from probe-by-probe global correlation analysis revealed remarkably uniform nucleosome occupancy signals across the 3-kb windows surveyed.
digested samples was greater than that of the heavily digested samples (blue arrows, Figure 2C). These regions, which were relatively more enriched in mononucleosomal DNA under the light-digest conditions than under the heavy-digest conditions, are interpreted as biochemically distinct footprints that are highly susceptible to MNase cleavage (Weiner et al., 2010; Henikoff et al., 2011; Xi et al., 2011) and will hereafter be called MNase-sensitive footprints (MSFs). We also observed the inverse pattern, in which nucleosome occupancy signals were greater under heavy-digest conditions than light-digestion conditions (red arrows, Figure 2C), indicative of footprints relatively resistant to being digested to nucleosomal monomers and therefore not detectable in the 100- to 200-bp fraction except under heavy-digest conditions. These regions exhibit persistent footprints even after extensive nuclease digestion, similar to those described in yeast (Xi et al., 2011), and are hereafter called MNase-resistant footprint (MRFs). The majority of regions that appeared hypervariable in the buffer comparison experiment frequently overlapped specifically with MSFs. This relationship was also observed when we averaged light, moderate, and heavy-digested nucleosome occupancy profiles at regions we identified as hypervariable in the buffer comparison experiment (SD > 1, n = 174; Figure 2D). These features were not a result of normalization artifacts because they are detectable in raw data and log-transformed data prior to normalization (Supplemental Figure 1). Thus, nucleosome occupancy mapping is susceptible to technical variability in the form of digestion-linked variation. The approach of mapping the nucleosome landscape with different degrees of digestion, or DNS-chip, was used to further explore the chromatin landscape in maize.

**MSFs and MRFs Are Nonrandomly Distributed within and Around Genes**

To expand the DNS-chip profiling beyond the 400 TSS regions to a larger and less biased sample of the maize genome, we designed a high-density microarray spanning three large blocks of the maize genome, each ;20 Mb. This "big-block" microarray includes 1688 of the 39,570 high-confidence genes (the maize "filtered gene set") in addition to single-copy intergenic regions. We performed DNS-chip with immature ears using the big-block microarray as shown in Figure 3. To quantify hypersensitivity and hyperresistance from DNS-chip, we subtracted the heavy-digest profile from the light-digest profile to create a "difference"
profile, where positive values in the indicate MNase sensitivity and negative values indicate MNase resistance (Figure 3A).

We computationally defined MSF and MRF segments at several statistical stringencies by applying a segmentation algorithm (iSeg; see Methods) to the difference profile data (example in segmentation track, Figure 3A, tabulated in Table 1). We found that MSFs and MRFs were not limited to TSS-proximal regions of known genes but were also observed within and between genes (Figure 3A). In particular, MSFs were frequently observed at the 5’ and 3’ boundaries of genes, whereas MRFs were frequently observed within the coding regions of genes (Figure 3A). Using the isSeg biological cutoff of 1.5 (see Methods and Table 1), we defined 3290 MSFs and 4908 MRFs within the 11 Mb of unique genomic space surveyed by the big-block microarray. A total of 58% of MSFs and 43% of MRFs were located in or near (within 500 bp) known protein-coding genes. Over half (58%) of all genes contained one or more MSFs. Similarly, most (67%) of all genes contained one or more MRFs. The MSFs had similar median segment size, 167 to 168 bp, in overall (n = 3290) and within-gene (n = 1602) regions. The MRFs had a median segment size of 115 bp in the overall region (n = 4908) region and 151 bp for those (n = 1815) found within genes. These findings show that MSFs and MRFs are widespread and abundant, occurring at most genes and representing previously unrecognized but major parts of the maize epigenomic landscape.

Aggregate DNS-chip profiles at the 1688 genes aligned by their boundaries revealed the presence of a prominent signal directly upstream of TSSs under light-digest conditions, at a location commonly referred to as the “nucleosome-depleted region” (arrow, Figure 3B). This signal is notably absent from both the moderate- and heavy-digest profiles. This finding is particularly important because the moderate- and heavy-digest conditions used here are typical of those used in published nucleosome mapping studies (Yuan et al., 2005; Johnson et al., 2006; Schones et al., 2008). Interestingly, this footprint signal is not predicted by a computational model of nucleosome occupancy likelihood trained on human chromatin (compared with Figures 3B to 3D; Fincher et al., 2013). Single gene examples (Figure 3A) and composite metagene analysis (Figure 3B) reveal a pattern in which many genes exhibit a tripartite chromatin structure in which MSFs are present at both ends of genes and MRFs reside within the open reading frames of genes. Thus, MSFs and MRFs show a nonrandom distribution in and around genes, which may indicate a function in gene regulation.

**MSFs Are Enriched at KNOTTED1 Binding Sites and Conserved Noncoding Sequences**

The prevalence and locations of MSFs around genes prompted us to ask whether these regions represent functional elements such as transcription factor binding sites. The binding sites for KNOTTED1 (KN1), a homeobox transcription factor expressed in meristematic cells (Vollbrecht et al., 1991), have been mapped via chromatin immunoprecipitation sequencing (ChIP-seq) in several tissues including immature ear (5 to 10 mm) and immature tassel (Bolduc et al., 2012). Although the immature ears we used for our study were harvested at a later stage of development (30 to 50 mm), we observed frequent overlaps between MSFs and KN1 binding sites (Figure 4). To examine the possible tissue specificity of MSF signatures, we classified kn1 ChIP-seq peaks into three exclusive groups: ear-specific (absent in tassel), tassel-specific (absent in ear), or common (present in both). On average, ear-specific kn1 ChIP-seq peaks displayed a conspicuous MSF signature, whereas tassel-specific kn1 ChIP-seq peaks showed only a slight MSF signature (Figure 4B). Interestingly, kn1 ChIP-seq peaks shared between ear and tassel displayed a very prominent MSF signature. We also examined the relationship between MSFs we identified in immature ears and ChIP-seq peaks identified in
maize pericarps of the MYB-domain-containing transcription factor PERICARP COLOR1 (P1) (Morohashi et al., 2012). P1, which shows little expression in maize ears compared with seeds (pericarp included) (Supplemental Figure 2), displayed average P1 ChIP-seq peaks that resembled an MRF-like signature (Figure 4C).

We next examined DNS-chip profiles at other potential regulatory elements defined as conserved noncoding sequences (CNSs; Turco et al., 2013). Like kn1 ChIP-seq peaks, CNSs also frequently overlapped with MSFs (Figure 4D). When we computed average DNS-chip profiles at CNSs aligned by their midpoint, we also found an MSF signature associated with CNSs (Figure 4E). Thus, CNS elements appear to exhibit a unique chromatin structural feature of an MSF. Together, these data show that many KN1 binding sites and CNSs in the maize genome can adopt a chromatin state that can be detected by DNS-chip, confirming the expectation that MSFs reflect functionally important genomic regions.

Promoter MSF Signals Are Associated with mRNA Levels and the Regulation of Gene Expression

To examine the relationship between MSFs and gene expression levels, we performed DNS profiling and RNA-seq on 9-d-old shoots. We then sorted the 1688 genes by their steady state mRNA levels and generated heat maps of DNS-chip profiles at the TSSs (Figures 5A and 5B). Gene expression levels were proportional to nucleosome occupancy signals under light-digest but not heavy-digest conditions. This trend was also observed when we grouped the 1688 genes into five equally sized groups sorted by expression levels and plotted average DNS-chip signals for each group (bottom plots, Figures 5A and 5B). When a heat map of the difference values (light-heavy) is generated, an expression-associated MSF signature (arrow, Figure 5C) is observed in all but the lowest quintile, a group with mostly nonexpressed genes (black line, Figure 5C). This analysis establishes that in general, the MSFs just upstream of TSSs are dynamic chromatin features, quantitatively captured by the DNS profiling method. Within gene bodies, expression levels were inversely proportional to nucleosome occupancy signals under both digestion conditions, a relationship consistent with that observed in other organisms (Valouev et al., 2011).

The correlation between gene expression and MSFs at TSSs suggests that differentially expressed genes between two given tissues may be associated with differences in the presence of MSFs within those genes. To further explore this relationship, we performed DNS-chip on root tissue from the same 9-d-old seedlings. Gene expression levels of each of the 1688 genes for the two organs were used to generate a shoot-versus-root scatterplot of gene expression values (Figure 5D). Each gene was grouped into expression tertiles for each tissue (Figure 5D, gray lines distinguishing groups S1-S3 and R1-R3). The root-versus-shoot expression-level scatterplot revealed that the majority of the 1688 genes represented on our microarray showed similar expression levels in the two organs. However, these groupings allowed us to examine a small subset of genes that were highly expressed in one tissue but showed little to no expression in the other. Examination of
representative genes whose RNA levels were high in shoot but low in root (S3R1) include those encoding chlorophyll A-B binding proteins, which contained an MSF in shoots but not roots (highlighted in yellow, Figure 5E). Conversely, genes whose RNA levels were low in shoot but high in root (S1R3) include those encoding an AT-hook motif Nuclear-localized protein22 and Nitrate transporter 2.4 (Figure 5F). These genes showed an MSF in roots but not shoots. Finally, genes whose RNA levels were high in both organs, such as Histone H3.2 and brittle stalk-2-like protein4, had pronounced MSF signals in both organs (Figure 5G).

Table 1. Summary Features of MSF and MRF Segments Defined at Different Statistical Stringencies by iSeg

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<th>Segment</th>
<th>SD Cutoff</th>
<th>Counts</th>
<th>Total Array (%)</th>
<th>Ave. No. of Segments per 10 kb</th>
<th>Segments Near Genes (%)</th>
<th>Segments within Genes (%)</th>
<th>Genes Near Segments (%)</th>
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MSF, MNase-sensitive footprint; MRF, MNase-resistant footprint.
SD cutoff: a parameter in the iSeg segmentation algorithm related to the statistical stringency of segmentation.
Counts: the total number of segments within the 11 Mb of unique genomic space surveyed by the big-block microarray.
Proportion: the sum of all segments divided by total array size (bp).
Near: within 500 bp.
Genes: The 1688 genes of the 39,570 canonical gene models (Maize Filtered Gene Set version 5b) represented on the microarray.

Figure 4. MSFs Overlap with Regulatory Elements.
(A) Nucleosome occupancy, Kn1 ChIP-seq peaks (Bolduc et al., 2012), the difference profile (light-heavy "difference"), and the segmentation of the difference profile ("segmentation") at a representative locus.
(B) Knotted1 ChIP-seq peaks were classified as tassel-specific, ear-specific, or shared between tissues, and each were aligned by their midpoints and used to calculate average nucleosome occupancy scores.
(C) Average nucleosome occupancy scores from developing ears at P1 ChIP-seq peaks identified in pericarps.
(D) Representative gene with data tracks for nucleosome occupancy, conserved noncoding sequences (Turco et al., 2013), the difference profile (light-heavy), and the segmentation of the difference profile.
(E) Average nucleosome occupancy scores at CNSs aligned by their midpoints.
to Supplemental Figure 3). As expected, nucleosome occupancy profiles based on heavy-digest data were highly correlated between the two tissues ($r = 0.92$). Interestingly, genes that were differentially expressed in root versus shoot did not on average display different levels of nucleosome occupancy based on heavy-digest data alone. Thus, the negative correlation between gene expression and nucleosome occupancy in the gene body is not reflected in genes that are differentially expressed between the two tissues (Supplemental Figure 3). Our findings show that MSFs can be tissue specific and that their presence, especially at the 5’ end of the genes, is coupled to mRNA expression levels. These observations, together with the colocalization of MSFs with CNSs and KN1 binding sites, demonstrate that DNS profiling is a powerful experimental approach for comprehensive mapping of functional chromatin landscapes in a plant genome.

**DISCUSSION**

In the process of developing a nucleosome occupancy mapping technique for maize, we found localized variability in nucleosome occupancy mapping that was likely caused by technical differences in the preparation of nucleosome-protected DNA. These variable regions were easily digested, or nuclease hypersensitive, resembling the “fragile nucleosomes” recently described for yeast (Weiner et al., 2010; Henikoff et al., 2011; Xi et al., 2011). We demonstrated the ability of DNS-chip to detect thousands of regions of fragile chromatin, which we termed MSFs, and identified their widespread presence and distribution in plants.

MSFs, experimentally defined as footprint signals evident under light-digest conditions that are less detectable under heavy-digest conditions, could have different causes (Figure 6). One possible biochemical basis for MSFs is a stable nucleosomal footprint that is preferentially released under light-digest conditions, which could result from a chromatin conformation that renders the surrounding linker DNA more susceptible to cleavage than that of most nucleosomes (Figure 6A). Nucleosomes that contain the histone variants H2A.Z and H3.3, which are more readily solubilized under moderate salt concentrations than bulk nucleosomes, are enriched at the “nucleosome-depleted” region in humans (Jin and Felsenfeld, 2007; Jin et al., 2009). Thus, the presence of MSFs in maize promoters may be a result of histone-variant-containing nucleosomes, which are conserved in maize. Given the association of MSFs with KN1 binding sites, another possible basis for MSFs could be non-nucleosomal footprints. Non-nucleosomal footprints such as those resulting from transcription factors could potentially yield a 100- to 200-bp fragment that may not persist under higher nuclease digest

**Figure 5. MSFs Are Associated with Gene Expression Levels.**

(A) to (C) Heat maps of nucleosome occupancy scores (light-digest and heavy-digest) and their difference (light-heavy) at 3 kb surrounding the 1688 TSSs included on the microarray. Gene TSSs were sorted based on their steady state gene expression levels (FPKM) and grouped into five equally sized groups (color coded at left of heat maps, from highest [red] to lowest [black]). Average scores for each of the five groups are plotted below each heat map. Blue arrow marks the -1 nucleosome region within which the difference profile covaries with gene expression levels.

(D) A scatterplot of root and shoot gene expression scores (log10 (1+FPKM)) for each of the 1688 genes. Genes were grouped into three equally sized tertiles based on their expression level for both root and shoot, demarcated by the gray lines, which generated nine groups of genes (S1-3, R1-3).

(E) to (G) Examples DNS-chip profiles for genes highly expressed in shoots but not in roots (S3R1), genes highly expressed in roots but not shoots (S1R3), and genes highly expressed in both tissues (S3R3). Examples include nucleosome occupancy profiles for each tissue along with RNA-seq read densities for each tissue.

(H) to (J) Average difference profiles (light-heavy) at all genes within the three groups aligned by their TSS, with the number of genes in each group indicated by “n=.”
levels (Figure 6B). In addition, it is possible that some MSFs may be caused by the simultaneous binding of transcription factors or chromatin remodelers and nucleosomes, which have been previously described (Adams and Workman, 1995; Floer et al., 2010; Mirny, 2010) (Figure 6C). In addition to nucleosome footprints that show different enrichment under different digestion conditions, there are also genomic regions that display a nucleosomal footprint under light-digestion conditions but display virtually no footprint under heavy-digest conditions (“Fragile,” Figure 6). These regions may represent a nucleosome conformation that renders the DNA on the nucleosomal core susceptible to cleavage with high concentrations of MNase, giving rise to <75-bp fragments that were not isolated when we gel-extracted 100- to 200-bp fragments.

Though MSFs may represent a variety of chromatin structures and footprints, we believe that most of the MSFs observed in this study result from biochemical properties of nucleosomes and not transcription factors. This expectation is largely based on the fact that these features are detected by the use of nucleosome-sized fragments (100 to 200 bp) on our microarrays and that transcription factor footprints are typically much smaller (Gidoni et al., 1985). This interpretation is also consistent with the observation that the majority of hypersensitive footprints identified in yeast overlap with signals of histone H3 (Xi et al., 2011). Though we do not have a maize DNase-hypersensitive site (DHS) data set to compare with, we believe that MSFs are distinct from DHSs. MSFs result from a pair of cleavages between 100 and 200 bp apart, in contrast to DHSs, which are defined by single cleavage events and are interpreted as nucleosome-depleted regions. It is conceivable that MSFs and DHSs may overlap at regions that possess appropriately spaced DHSs in individual cells, though the relationship between fragile nucleosomes and DHSs has not been examined.

Nucleosome mapping studies in a variety of model systems have reported the pervasive presence of nucleosome-depleted regions directly upstream of TSSs (Lee et al., 2007; Schones et al., 2008; Valouev et al., 2008, 2011). In humans, many so-called nucleosome-depleted regions in promoters overlap with H2AZ/H3.3-containing nucleosomes located just upstream of TSSs, and these nucleosomes are more readily solubilized in moderate-salt-containing buffers (Jin and Felsenfeld, 2007; Jin et al., 2009). This suggests that some regions that have been referred to as nucleosome-depleted may in fact be nucleosomal footprints that are not efficiently recovered under conditions commonly used to prepare nucleosomal DNA. In this study, we showed that occupancy footprints are readily detected in the nucleosome-depleted region of many genes under light digest conditions, supporting the idea that some regions classified as “nucleosome-free” may in fact represent fragile nucleosomes.

Another pattern revealed in our study is the presence of genomic regions whose relative abundance increased with higher levels of nuclease digestion (MRFs, Figure 6). This pattern could result from tightly packed arrays of nucleosomes with shorter or protected linker DNA, rendering the nucleosomes relatively resistant to MNase cleavage. Whatever the cause, this pattern is typical of open reading frames (e.g., the *Bronze1* open reading frame region [Figure 3A] and the trend plot in Figure 3B). The localization of MRFs within gene bodies, particularly directly downstream of the TSS, indicates that they may be involved in transcriptional elongation, but their importance remains to be characterized.

DNS profiling is an approach to chromatin profiling that reveals multiple dimensions of chromatin structure information, including nucleosome occupancy, nuclease sensitivity, and chromatin compaction. In this study, we identified the presence of thousands of hypersensitive footprints in a plant genome that are associated with multiple different features, including transcription factor binding.

![Figure 6. Potential Causes of MSFs and MRFs.](image-url)

A hypothetical set of nucleosome occupancy profiles using light and moderate digestion conditions, with “Open” MSF, “Fragile” MSF, and MRF signatures (top). Footprints that could potentially give rise to the observed nucleosome occupancy profiles are shown. Regions that are preferentially cut under light-digestion conditions are indicated by blue lightning bolts, whereas regions that are preferentially cut under heavy-digest conditions are indicated by red lightning bolts. An “Open” MSF signature could potentially be caused by a nucleosomal footprint (A), a transcription factor footprint (B), or the simultaneous binding of both (C) that render the surrounding DNA hypersensitive to MNase digestion. A “Fragile” MSF signature could potentially be caused by a nucleosomal footprint (A) or transcription factor footprint (B), or the simultaneous binding of both (C) that are susceptible to over digestion by MNase. MRFs could potentially be caused by arrays of tightly packed nucleosomes (A) to (C).
sites, conserved noncoding sequences, gene expression levels, and tissue-specific gene expression. This approach has great potential as a framework to identify functionally important regions of the genome and to integrate different types of genomic annotations.

**METHODS**

**Plant Material**

Maize (Zea mays) B73 seeds were obtained from the Maize Genetics Cooperation Stock Center. To obtain seedling tissue, seeds were germinated in Fafard Seedling Mix in the greenhouse (Department of Biological Science, Florida State University, Tallahassee, FL). At midday, 9 d after germination, aboveground tissue was cut at the soil line, flash frozen in liquid nitrogen, and stored at −80°C until use. Immature ears 3 to 5 cm in size were obtained from field-grown plants at the Florida State University Mission Road Research Facility (Tallahassee, FL).

**Nucleus Isolation Buffer Compositions**

APEL (20 mM Tris HCl, pH 7.8, 250 mM sucrose, 5 mM MgCl₂, 5 mM KCl, 40% glycerol, 0.25% Triton X-100, and 0.1% BME; Steinmüller and Apel, 1986), BFA (15 mM PIPES-NaOH, pH 6.8, 0.32 mM sorbitol, 80 mM KCl, 20 mM NaCl, 0.5 mM EGTA, 1 mM EDTA, 0.15 mM spermine, and 0.5 mM spermidine; Belmont et al., 1987), and MNEP (15 mM Tris-HCl, pH 7.5, 300 mM sucrose, 5 mM MgCl₂, 60 mM KCl, 15 mM NaCl, 0.1 mM phenylmethylsulfon fluoride, 0.1 mM EDTA, 0.5 mM DTT, and 1% Triton X-100; Dennis et al., 2007) were used for buffer comparison experiments. All other experiments used BFA for nucleus isolation. Immediately before use, 10 mM o-phenanthroline and 0.1 mM phenylmethylsulfon fluoride were added to nuclear fixation buffers.

**Nucleus Isolation**

Ten grams of seedling tissue, or 1 g of immature ear tissue, were ground under liquid nitrogen with a mortar and pestle and cross-linked by stirring for 10 min in 10 mL ice-cold nuclear isolation buffer with 1% formaldehyde. Fixation was stopped with 125 mM glycine for 5 min. Tissue was pelleted at 2000g for 10 min at 4°C in a swinging-bucket centrifuge, then the tissue pellet was re-suspended in 25 mL nucleus isolation buffer with 1% hexylene glycol and 1% Triton X-100. After 5 min of gentle rotatory shaking, nuclear suspensions were filtered through two layers of Miracloth (Calbiochem) and nuclei were pelleted at 2000g for 10 min at 4°C. Nuclei were washed once with 10 mL nuclear isolation buffer and re-suspended in 10 mL Nase digestion buffer (50 mM Tris-HCl, pH 7.5, 320 mM sucrose, 4 mM MgCl₂, and 1 mM CaCl₂). Nuclei were sequentially filtered through 150, 100, 50, and 30 µm CellTric filters (Partec), pelleted at 2000g for 10 min at 4°C, resuspended in 1 mL of MNase digestion buffer, flash frozen in liquid nitrogen, and stored at −80°C until use.

**Cytology and Quantitation of Purified Nuclei**

To assess nuclear preservation, nuclei were placed on standard glass microscopy slides, stained with 3 µg/mL 4’,6-diamidino-2-phenylindole in 1× PBS, and mounted in Vectashield antifading reagent with a nail-polish-sealed 22 × 30 × 1.5-mm cover slip. Images were collected on an Olympus IX-70 epifluorescence microscope and deconvolved and analyzed with the SoftWorx computerized workstation.

**Isolation of Nucleosomal DNA**

To compare the effect of nuclear isolation buffer on nucleosome occupancy, 100 µL nuclei were digested with 2 units/mL MNase at 37°C for 5 min and stopped with 20 mM EGTA. Deionized H₂O (100 µL) was added to each sample, and nuclei were de-cross-linked with 1% SDS and 100 µg/mL protease K overnight at 65°C. DNA was phenol-chloroform extracted, ethanol precipitated, and resuspended in deionized water. Digested DNA were treated with 40 µg/mL RNase A and run in a 1% agarose gel. DNA fragments (100 to 200 bp) were excised and gel extracted with the Qiagen II gel extraction kit (Qiagen) following the manufacturer’s instructions.

To examine the effect of digestion level on nucleosome occupancy in immature ears, nuclei isolated with the three buffers (APEL, HNIB, and BFA) were pooled, and 500 µL nuclear suspensions was digested with 2 units/mL MNase (light), 30 units/mL MNase (moderate), or 300 units/mL MNase (heavy). A total of four replicates were performed for each digestion level; however, replicate 4 of the light digest was discarded because of an anomalous microarray probe score distribution.

**RNA-seq Library Preparation and Data Analysis**

RNA was prepared from the same pools of seedling shoot and root tissue used for DNS-seq experiments. One hundred milligrams of frozen, ground tissue was resuspended in 1.5 mL Trizol (Invitrogen) and prepared according to the manufacturer’s instructions. RNA-seq data were kindly provided by Cornell University Genomics Center (courtesy of E.S. Buckler, E. Rodgers-Melnick, and P.A. Schweitzer) using RNA from our samples. The RNA-seq libraries were prepared using the Illumina TruSeq RNA sample preparation kit (v2) according to the manufacturer’s instructions. RNA-seq libraries were subjected to 100-cycle, single-end sequencing with an Illumina HiSeqation 2500. Single-end reads were aligned to the AGPv2 genome assembly with TopHat (Kim et al., 2013) using default parameters. Fragments per kilobase per million (FPKM) values for each gene were obtained with Cutffind (Trapnell et al., 2013) using default parameters.

**Microarray Design and Hybridization**

The NimbleGen microarrays contain isothermal 49- to 74-bp probes at ~15-bp spacing, representing both strands of genomic sequence. The 12-plex TSS-based microarray covered 3000 bp of genomic sequence surrounding 400 transcription start sites primarily derived from the Classical Maize Genes (Schnable et al., 2009). The 3-plex “big-block” NimbleGen microarrays represent unique regions within three ~20-Mb gene-rich chromosome termini of chromosomes 4L, 5S, and 9S. Purified nucleosomal DNA and B73 bare genomic DNA were labeled and hybridized to the microarray according to manufacturer’s instructions for Dual-color CGH microarrays (Nimblegen).

**Data Analysis and Sources**

Slides were scanned with a Nimblgen MS 200 Scanner, and probe-by-probe fluorescence intensity ratios (nucleosomal/genomic) were produced with Nimblescan 2.6. Subsequent data analysis and plotting were performed primarily using BEDtools (Quinlan and Hall, 2010) and the R statistical programming language (Ihaka and Gentleman, 1996). Nucleosome occupancy profiles at each base pair were calculated as the average score of all overlapping probes at each base pair. Profiles were then smoothed using a 60-bp sliding-mean window at 10-bp step sizes. Data for each sample were quantile normalized. Segmentation of nucleosome occupancy and difference profiles was performed by first discarding regions with probes covering fewer than 500 bp of contiguous sequence, allowing for small (~60 bp) gaps. The resulting data were then used for segmentation with iSeg (http://cloud.stat.fsu.edu/iSeg). Segments exceeding a 20% cutoff of 1.5 were used for “segmentation” tracks and overlap quantifications. Segmentation tracks for each profile will be available on the Greenome browser, a new UCSC genome browser (Kent et al., 2002) for plant genomes (http://greenomebrowser.com/). ChiP-seq peaks were obtained from the Gene Expression Omnibus (GEO; http://www.ncbi.nlm.nih.gov/geo/) for km1 (GSE39161; Bolduc et al., 2012) and p1 (GSE30897; Morohashi et al., 2012). Canonical gene models
were obtained from the B73_RefGen_v2 5b version of Filtered Gene Set (ftp://ftp.ensemblgenomes.org/pub/plants/release-15/gtf/zea_mays/). p1 gene expression profiles were obtained from qTeller (qTeller.org) using published RNA-seq data (Jia et al., 2009; Wang et al., 2009; Li et al., 2010; Davidson et al., 2011; Waters et al., 2011).

Accession Numbers
Microarray data have been deposited in the National Center for Biotechnology Information GEO database under accession number GSE60092. ChIP-seq peaks were obtained from GEO for knr1 (GSE39161) and p1 (GSE38587).

Supplemental Data
The following materials are available in the online version of this article.

Supplemental Figure 1. MNase-Sensitive and MNase-Resistant Footprints Are Observed in Raw and Un-Normalized Data.

Supplemental Figure 2. PERICARP COLOR1 Gene Expression Levels in Various Maize Tissues.

Supplemental Figure 3. Average Nucleosome Occupancy and Difference Profiles at Genes Grouped by Their Expression Levels in Seedling Shoot and Roots.

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AUTHOR CONTRIBUTIONS

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