Supplemental Data

Variation in Homeodomain DNA Binding

Revealed by High-Resolution Analysis

of Sequence Preferences

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1. Cloning

Cloning homeodomains into pMAGIC1

Homeodomain open reading frames, consisting of the pfam-defined homeodomain and 15 amino acids of flanking sequence (or to the end of the full open reading frame) were cloned into pMAGIC1 (1) by either RT-PCR from pooled mouse mRNA (2) followed by ligation independent cloning, or by gene synthesis (DNA 2.0) followed by conventional cloning using BamHI and NotI restriction sites. All clones were sequence verified in pMAGIC1 and are given in the file "Protein and DNA sequence". The inserts were then transferred into a T7-GST-tagged variant of pML280 according to protocols described in (1). The resulting recipient plasmids after transfer express N-terminal GST fusion proteins fused to the DBD flanked by H3 and H4 regions used in the recombination step (**bold**):

MSPILGYWKIKGLVQPTRLLLEYLEEKYEEHLYERDEGDKWRNKKFELGLEFPNLPYYI DGDVKLTQSMAIIRYIADKHNMLGGCPKERAEISMLEGAVLDIRYGVSRIAYSKDFETLK VDFLSKLPEMLKMFEDRLCHKTYLNGDHVTHPDFMLYDALDVVLYMDPMCLDAFPKL VCFKKRIEAIPQIDKYLKSSKYIAWPLQGWQATFGGGDHPPKSDLVPRPCEL**KLDVHML VPRGSLEVLFQGPEGDATMGHMVHRPWIQ** – DBD region -**AWPQGGRTRIVSAHNSENLYFQGDLRGSITN** GSGC*

2. Protein production and quantization

We produced proteins by two methods that yielded essentially identical results: Expression and purification from *E. coli*, and expression by *in vitro* transcription/translation (IVT). (See next page for a graph comparing results from the two systems.)

Expression and purification in E. coli. We transformed homeodomain-encoding constructs into *E. coli* C41 DE3 cells (Lucigen). Freshly-transformed cultures were grown overnight in LB medium containing 50 mg/ml ampicillin and diluted 1:100 in fresh LB medium. The cells were grown at 25°C to a final concentration of $OD_{600} \sim 0.8$ and then induced with IPTG (Bioshop) to a final concentration of 1 mM. These cultures were then grown overnight at 14°C. Cell pellets were obtained by centrifugation at 4°C for 15 minutes at 4000 rpm. Each pellet was resuspended in cold lysis buffer (50mM Tris pH 8, 150 mM NaCl, 2mM DTT, and 12.8 mg of lysozyme). The resuspension was incubated in ice for 20 minutes and lysed by sonication. Cell lysates were centrifuged at 4°C for 15 minutes at 4000 rpm and soluble fraction was collected. GST resin slurry (Amersham) was added to the fraction and binding proceeded at 4°C for 45 minutes. The beads were washed 2-3 times with PBS with 2 mM DTT and then incubated with elution buffer (50 mM Tris pH 7.5, 10 mM reduced glutathione, Roche protease inhibition and 2 mM DTT) at 4°C for 1 hr. Concentration of GST-tagged DBD was estimated for each protein relative to a dilution series of GST standards on Coomassie-stained SDS-PAGE gels.

In vitro translation. For *in vitro* translation reactions, the manufacturer's protocol (Ambion ActivePro Kit) was followed. The molarities of all *in vitro* translated proteins were determined by Western blot using a dilution series of recombinant GST (Sigma). Equal volumes of sample and known concentrations of GST were suspended in 1x NuPAGE LDS Sample Buffer (Invitrogen), heated to 95°C for 5 minutes, and loaded on a Bio-Rad 4-12% Bis-Tris Criterion precast gel (Bio-Rad). Samples were electrophoresed at 200 V for 25 minutes and then transferred to a nitrocellulose membrane (Sigma) at 30 V for 3 hours. Membranes were labeled and developed using the SuperSignal West Femto Maximum Sensitivity Substrate kit (Pierce) according to the manufacturer's protocols. Primary antibody (anti-GST produced in rabbit; Sigma) was added at 20 ng/ml, and secondary antibody (horseradish-peroxidase-conjugated anti-rabbit IgG produced in goat; Pierce) was added at 5 ng/ml. Film was scanned and analyzed using Quantity One version 4.5.0 software (Bio-Rad) according to the GST standard curve.

Glycerol was added to a final concentration of 30% to both IVT and purified protein samples prior to storage.

Figure S1. The plot below shows results from Lhx6, assayed as an IVT protein from one plasmid, and as a protein expressed in E. coli and purified, from a different construct, illustrating the reproducibility of the system and robustness to protein production method:



3. Microarray methods

Microarray Design

The construction of 'all 10-mer' universal protein binding microarrays (PBMs) using a de Bruijn sequence of order 10 has already been described (3) and is described in more detail in conference proceedings posted at http://thebrain.bwh.harvard.edu/RECOMB2007.pdf. For this study, we further optimized our design to achieve greater coverage of gapped *k*-mers, as described below. A de Bruijn sequence of order *k* is a circular string of length 4^k that contains every *k*-mer exactly once when overlaps are considered. To generate a de Bruijn sequence of order 10 for our universal PBM, we used a linear-feedback shift register corresponding to the primitive polynomial:

 $3x^{10} + 3x^{9} + 2x^{8} + 1x^{7} + 2x^{6} + 2x^{5} + 3x^{4} + 3x^{3} + 1x^{2} + 2x$

We empirically selected this particular de Bruijn sequence because it uniformly covers all contiguous 10-mers and all gapped 10-mers spanning 11 total positions. Further, it exhibits optimal coverage of contiguous and gapped 8-mers. Any 8-mer is guaranteed to occur 16 times in our de Bruijn sequence (32 times for non-palindromes). Our de Bruijn sequence exhibits this 16/32-fold redundancy for all gapped 8-mers spanning up to 12 total positions (except for sequence variants of the single pattern 1111-1-1-1), as well as all gapped 8-mers of the pattern 1111-gap-1111 with a gap of up to 20 positions. Thus, all 4⁸ sequence variants for each of these 341 patterns (more than 22.3 million 8-mers) occur at least 16 times each.

After generating this de Bruijn sequence *in silico*, we partitioned it into subsequences of length 36 nucleotides (nt) and overlapping by 11 nt, resulting in 41,944 36-mers. Any 36-mer with a run of five or more consecutive guanines was replaced by its reverse complement to avoid problems in double-stranding (see below). We appended a common 24-nt sequence to each 3' end (5'-gtctgtgttccgttgtccgtgctg-3') complementary to our primer for double-stranding (5'-cagcacggacaacggacaacggacacagac-3') in order to create 60-mer sequences that would become the probes on our custom-designed microarray. These microarrays were synthesized by Agilent technologies in their "4x44K" format, with probes attached to the glass slide at the 3' end. Each slide contains the entire complement of all possible 10-mers in four identical subgrids of approximately 44,000 probes each, which can be physically separated into four chambers for four separate experiments. The additional probes beyond the set of 41,944 were designated as control sequences for a variety of purposes. All microarray probe sequences used in this study are listed on our website, <u>http://the_brain.bwh.harvard.edu/pbms/webworks2/</u>.

Protein Binding Microarrays

Protein binding microarray (PBM) experiments were performed essentially as described previously, except that four proteins were simultaneously assayed in separate sectors of a single microarray (3). First, single-stranded oligonucleotide microarrays were double-stranded by primer extension and scanned on a microarray scanner (GSI Lumonics ScanArray 5000) prior to protein incubation. Primer extension reactions consisted of 1.17 µM HPLC-purified common primer (Integrated DNA Technologies), 40 µM dATP, dCTP, dGTP, and dTTP (GE Healthcare), 1.6 µM Cy3 dUTP (GE Healthcare), 32 Units Thermo Sequenase[™] DNA Polymerase (USB), and 90 µl 10x reaction buffer (260 mM Tris-HCl, pH 9.5, 65 mM MgCl₂) in a total volume of 900 µl. The reaction mixture, microarray, stainless steel hybridization chamber, and singlechamber gasket cover slip (Agilent) were pre-warmed to 85°C in a stationary hybridization oven and assembled according to the manufacturer's protocols. After a two-hour incubation (85°C for 10 min, 75°C for 10 min, 65°C for 10 min, and 60°C for 90 min), the hybridization chamber was disassembled in a glass staining dish in 500 ml phosphate buffered saline (PBS) / 0.01% Triton X-100 at 37°C. The microarray was transferred to a fresh staining dish, washed for 10 min in PBS / 0.01% Triton X-100 at 37°C with a magnetic stir bar, washed once more for 3 min in PBS at 20°C, and spun dry by centrifugation at 40 g for 1 min.

For protein binding reactions, double-stranded microarrays were first pre-moistened in PBS / 0.01% Triton X-100 for 5 min and blocked with PBS / 2% (wt/vol) nonfat dried milk (Sigma) under LifterSlip cover slips (Erie Scientific) for 1 h. Microarrays were then washed once with PBS / 0.1% (vol/vol) Tween-20 for 5 min and once with PBS / 0.01% Triton X-100 for 2 min. Proteins were diluted to 100 nM (unless otherwise specified) in a 175-µl protein binding reaction containing PBS / 2% (wt/vol) milk / 51.3 ng/µl salmon testes DNA (Sigma) / 0.2 µg/µl bovine serum albumin (New England Biolabs). Preincubated protein binding mixtures were applied to individual chambers of a four-chamber gasket cover slip in a steel hybridization chamber (Agilent), and the assembled microarrays were incubated for 1 h at 20°C. Microarrays were again washed once with PBS / 0.5% (vol/vol) Tween-20 for 3 min, and then once with PBS / 0.01% Triton X-100 for 2 min. Alexa488-conjugated rabbit polyclonal antibody to GST (Invitrogen) was diluted to 50 µg/ml in PBS / 2% milk and applied to a single-chamber gasket cover slip (Agilent), and the assembled microarrays were again incubated for 1 h at 20°C. Finally, microarrays were washed twice with PBS / 0.05% (vol/vol) Tween-20 for 3 min each, and once in PBS for 2 min. Slides were spun dry by centrifugation at 40 g for 5 min. After each hour-long incubation step, microarrays and cover slips were disassembled in a staining dish filled with 500 ml of the first wash solution. All washes were performed in Coplin jars at 20°C on an orbital shaker at 125 r.p.m. Immediately following each series of washes, microarrays were rinsed in PBS (slowly removed over approximately 10 seconds) to ensure removal of detergent and uniform drying.

Microarray Stripping

Protein and antibody were digested from double-stranded microarrays in a 70-ml stripping solution consisting of 10 mM EDTA, 10% SDS, and 290 Units of protease (from *Streptomyces griseus*; Sigma), shaking at 200 r.p.m. in a Coplin jar at 37°C for 16 hours. Microarrays were then washed 3 times for 5 minutes each in PBS / 0.5% (vol/vol) Tween-20, once for 5 minutes in PBS, and finally rinsed in PBS in a 500-ml staining dish (slowly removed over approximately 10 seconds) to ensure removal of detergent and uniform drying. All washes were performed in Coplin jars at 20°C on an orbital shaker at 125 r.p.m. Before re-use, slides were scanned once at the highest laser power for Alexa488 (488 nm excitation (ex), 522 nm emission (em)) to ensure that no protein or antibody signal remained, and once for Cy3 (543 nm ex, 570 nm em) to ensure that there was no appreciable loss in DNA quantity. For this study, all PBM experiments were performed either on a fresh slide or a slide that had been stripped exactly once, which yielded indistinguishable results (data not shown).

Image Quantification and Data Normalization

Protein-bound microarrays were scanned to detect Alexa488-conjugated antibody (488 nm ex, 522 nm em) using at least three different laser power settings to best capture a broad range of signal intensities and ensure signal intensities below saturation for all spots. Separately, slides were scanned after primer extension to quantify the amount of incorporated Cy3-conjugated dUTP (543 nm ex, 570 nm em). Microarray TIF images were analyzed using GenePix Pro version 6.0 software (Molecular Devices), bad spots were manually flagged and removed, and data from multiple Alexa488 scans of the same slide were combined using masliner (MicroArray LINEar Regression) software (4).

Our two-step method of raw data normalization was described previously (3). First, we normalize Alexa488 signal by the Cy3 signal for each spot to account for differences in the total amount of double-stranded DNA. Because Cy3-dUTP incorporation is influenced both by the total number of adenines and the sequence context of each adenine, we perform a linear regression over all 41,944 variable spots to compute the relative contributions to the total signal of all trinucleotide combinations (followed by adenine). Using these regression coefficients, we calculate the ratio of observed-to-expected Cy3 intensity and use that as a normalization factor. Second, to correct for any possible non-uniformities in protein binding, we further adjust the Cy3-normalized Alexa488 signals according to their positions on the microarray. We calculate the median normalized intensity of the 15 x 15 block centered on each spot and divide the spot's signal by the ratio of the median within the block to the median over the entire chamber. Raw and normalized forms of the data for all experiments in this study are provided on our supplementary website, <u>http://the_brain.bwh.harvard.edu/pbms/webworks2/</u>.

4. Data representation

Sequence Analysis and Motif Construction

Due to the 32-fold redundancy of 8-mers described above (i.e., every non-palindromic 8-mer occurs on at least 32 spots in each chamber of our universal PBM), we are able to provide a robust estimate of the relative preference of a transcription factor for every contiguous and gapped 8-mer that is covered on our array. Here, we provide several scores for each 8-mer: (1) Median Intensity, (2) Z-Score, (3) Enrichment Score (E-Score), and (4) False Discovery Rate Q-Value for the E-Score.

Median intensity refers to the median normalized signal intensity for the set of \sim 32 probes harboring a match to each 8-mer. We have previously shown that PBM median signal intensity is proportional to binding affinity (3). The distribution of log (median intensity) over all 8-mers is used to compute a Z-Score for each 8-mer according to the following formula:

$$Z-Score = \frac{\log \text{ median intensity of kmer} - \log \text{ median intensity of all kmers}}{\text{robust estimate of standard deviation}}$$

Here, our robust estimate of the standard deviation is the median absolute deviation (MAD), multiplied by 1.4826 for normally distributed data (5). (We have observed, though, that the distribution of the log median intensity is not a normal distribution. Therefore, the inferred Z-Score should not be mistaken for the usual Z-Score annotation in the literature.) Both the median intensity and Z-Score are advantageous because they retain information regarding relative differences in signal intensity, and thus probe occupancy and relative affinity as well. However, experimental variability and differences in absolute signal intensity and non-specific binding can make these measures difficult to compare for different transcription factors.

Our E-Score is a rank-based, non-parametric statistical measure that is invariant to protein concentration and readily allows different experiments to be compared on the same scale. This score has already been described in detail (*3*). Briefly, for each 8-mer (contiguous or gapped) we consider the collection of all probes harboring a match as the "foreground" feature set and the remaining probes as a "background" feature set. We compare the ranks of the top half of the foreground with the ranks of the top half of the background by computing a modified form of the Wilcoxon-Mann-Whitney (WMW) statistic scaled to be invariant of foreground and background sample sizes. The E-Score ranges from +0.5 (most favored) to -0.5 (most disfavored). Finally, we compute a False Discovery Rate Q-Value for the E-Score by comparing it to the null distribution of E-Scores (over 32,896 8-mers) calculated by randomly shuffling the mapping among the 41,944 probe sequences and intensities (repeated 20 times) (6) We note that in

computing all of the above scores, we do not consider probes for which the 8-mer occupies the most distal position on the probe (5' with respect to the template strand) or for which the 8-mer overlaps the 24-nt primer region.

Figure S2. Shown below is a scatter plot comparing E and Z for four homeodomains (Hoxa3, Tcf1, Six6, Nkx3-1) illustrating how the two measures relate to one another. In essence, the E-score and Z-score capture essentially the same information, but the E-score representation is flattened as values approach 0.5.



In addition to reporting scores for each individual 8-mer, we compactly represent these data in a position weight matrix (PWM) for each TF using our "Seed-and-Wobble" algorithm, which has been described previously (3). The algorithm works in two stages. In the first "Seed" stage, we identify the single 8-mer (contiguous or gapped) with the greatest enrichment score. For this study, we considered all 8-mers spanning up to 10 total positions as candidate seeds. In the second "Wobble" stage, we systematically test the relative preference of each nucleotide variant at each position, both within and outside the seed. This is accomplished by examining each of the four nucleotides at each position within the 8-mer seed (keeping the other 7 positions fixed) and computing the modified WMW statistic using the entire set of probes containing one of the four variants. For positions outside the 8-mer seed, we first identify the single position within the seed with the lowest information content, treat it as a gapped position, and query every other

position outside the seed for which the resulting 8-mer is covered in our de Bruijn sequence (i.e., all 4^8 sequence variants of that pattern exhibit 32-fold redundancy). Finally, we transform the motif derived from this method into a PWM using Boltzmann distribution. Importantly, this method takes advantage of the fact that all sequence variants occur an equal number of times on the microarray, and it considers all features without applying any arbitrary cutoffs.

In the analysis below, we applied the Seed-and-Wobble algorithm to the construction of "TF-specific" motifs that capture the binding sites that are preferred by one TF relative to another TF or group of TFs. To derive a TF-specific PWM for one factor (TF_A) relative to a single other factor (TF_B), we first rank all microarray probes according to the ratio of their ranks for the two experiments (i.e., rank TF_A / rank TF_B). We then identify the 8-mer with the highest enrichment score in the new probe ranking and use Seed-and-Wobble to generate a TF-specific PWM in the new ranking. To derive a TF-specific PWM for one factor (TF_A) relative to a group of other factors (TF_{group}), our approach is similar except that we rank probes based on the average rank for the entire group (i.e., rank TF_A / rank (avg rank TF_{group})). In both cases, we limit to only those probes for which we have data in all experiments under consideration.

In order to establish which representation should be used as a standard, we replicated four experiments on a second array with completely different probe sequences (but still containing all possible 10-mers). We found that 8-mer E- and Z-scores determined from one array were substantially better predictors of the 8-mer E- and Z-scores on the other array, in comparison to the relative preference of each 8-mer predicted from the dominant motif obtained from the data on the other array.

Moreover, E-scores were typically a slightly better predictor than Z-scores. The following graphs show Precision at 70% Recall. Positives were defined by a moving threshold on E-scores (or Z-scores, if Z-scores are used as training/test) for the second independent array. At each threshold (i.e. each point on the graph) the precision statistic (True Positives / (True Positives + False Positives)) was determined for the value at which the recall statistic (True Positives / (True Positives + False Negatives)) is 70%. E-score (red), Z-score (blue), and SW (Seed-and-Wobble) (*3*) motifs (black and green) were defined on the original array. For the precision-recall analysis, motif scores were calculated for all 8-mers using the Gomer scoring method (7), which is an estimate of the probability of transcription factor binding. These graphs demonstrate the reproducibility in our measurements for separate 8-mers and emphasize the importance of retaining individual scores for all 8-mers rather than compressing these data into a motif which cannot fully recapitulate these preferences.

Figure S3. Results from training on the version of the array used for all homeodomains, and testing on an alternative array design.



Figure S4. Results from training on the alternative array design, and testing on the version used for all homeodomains:



5. PBM motifs and comparison to motifs in Transfac and Jaspar

Table S1. Logos constructed using our Seed-and-Wobble algorithm are shown for all proteins presented in this study. When available, the corresponding position weight matrix (PWM) assembled from the literature is also shown for the mouse protein or its closest ortholog in any other metazoan species. PWMs were retrieved from the JASPAR (8) and TRANSFAC (9) databases, as indicated. These PWMs were mainly derived either from *in vitro* selection (Selex) experiments or from a compilation of validated binding sites from a variety of sources and experimental methods. Of the 168 mouse proteins examined here, 6 had a PWM in the JASPAR database (10 to 59 sequences), 28 more had a PWM in the TRANSFAC database (5 to 56 sequences), and 23 had between 1 and 5 individual binding sequences listed. For the remaining 111 proteins, we searched for additional binding data corresponding to the closest ortholog of each protein in human, zebrafish, D. melanogaster, C. elegans, and sea urchin. The ortholog was considered to be the best ungapped BLAST match in the other species over at least 35 AAs. We identified 40 proteins possessing an ortholog with a reported binding matrix or sequence, bringing our total to 97 proteins with binding data. In cases where binding data existed for both the mouse protein and an ortholog, we considered the more comprehensive data set. (h = human; m = mouse; r = rat).

TF Name	PBM Seed-and-Wobble	Best Hit	Database	Database PWM
Alx3				
Alx4		Alx4 (h,m)	TRANSAFAC 5 compiled seq.	GATTATTCTCAGG
Arx	TAATTAA			
Bapx1	CACTTA	Bapx1 (m)	JASPAR 24 Selex seq.	
Barhl1	AAGCAATTA			
Barhl2	A CAATTAA			
Barx1	TAT SA			
Barx2	TAATTA			
Bsx	TAATTA A			
Cart1	TAATTA	Cart1 (h)	TRANSFAC 25 Selex seq.	

Cdx1	GIAA AAA	Cdx1 (h,m)	TRANSFAC 13 compiled seq.	A
Cdx2	AAATAS	Cdx2 (m)	TRANSFAC 9 compiled seq.	SE ALAAA I I
Cphx	TGAT ATCA	Eve (fly)	TRANSFAC 9 compiled seq.	GC G SET A
Crx	T THE GOATTA SE	Crx (h,m,r)	TRANSFAC 26 compiled seq.	ATTA ^
Cutl1	APTATA	Cutl1 (h)	TRANSFAC 86 Selex seq.	T_ATCAT
Dbx1	A-1+0ATTA+1+-+			
Dbx2	ATTAATIA_I.			
Dlx1	TAATTA _			
Dlx2	- AATTA			
Dlx3	TAATTA			
Dlx4				
Dlx5				
Dmbx1	EGATTA			
Dobox4	TAGATACCC			
Dobox5	GGATTAA			
Duxl				
Emx2	TAATTA			
En1	AA TAATTA	En1 (m)	JASPAR 10 Selex seq.	
En2		En (fly)	TRANSFAC 11 compiled seq.	TAATIG
Esx1	TAATTA			
Evx1		Eve (fly)	TRANSFAC 9 compiled seq.	GC G SET A
Evx2		Eve (fly)	TRANSFAC 9 compiled seq.	GC G SET A

Gbx1				
Gbx2				
Gsc	GGATTA			
Gsh2				
Hdx				
Hlx1	TAATTAATTA			
Hlxb9	TAATTA			
Hmbox1	TAACIAG			
Hmx1				
Hmx2				
Hmx3		Hmx3 (m)	TRANSFAC 11 Selex seq.	CAAGTGCGTG
Homez				
Hoxa1	TAATTA			
Hoxa2	TAATTA			
Hoxa3	TAATTA	Hoxa3 (m)	TRANSFAC 14 Selex seq.	c_TA
Hoxa4	TAATTA	Hoxa4 (h,m)	TRANSFAC 6 compiled seq.	ATTEG
Hoxa5	TAATTA	Hoxa5 (m)	TRANSFAC 10 compiled seq.	an ATATALA. I an
Hoxa6	TAATTAGE &	Antp (fly)	TRANSFAC 13 compiled seq.	AATAASAATAA
Hoxa7	TAATTA	Hoxa7 (h,m)	TRANSFAC 18 Selex seq.	-CAATC I
Hoxa9	Sec ATAAAA	Abd-B (fly)	TRANSFAC 45 Selex seq.	
Hoxa10	STAAT AAA			
Hoxa11	GTCGTAAA			

Hoxa13	CTCSTAAA			
Hoxb3	TATA	Dfd (fly)	TRANSFAC 57 selected seq.	
Hoxb4	TAATTAA	Dfd (fly)	TRANSFAC 57 selected seq.	
Hoxb5	TAATTA .			
Hoxb6	TAT STATTAC	Antp (fly)	TRANSFAC 13 compiled seq.	AATA
Hoxb7		Antp (fly)	TRANSFAC 13 compiled seq.	AATA
Hoxb8	SeaATTAA	Antp (fly)	TRANSFAC 13 compiled seq.	AATA
Hoxb9	AAAA	Antp (fly)	TRANSFAC 13 compiled seq.	AATA
Hoxb13	- CEATAAAA	Abd-B (fly)	TRANSFAC 45 Selex seq.	
Hoxc4		Dfd (fly)	TRANSFAC 57 selected seq.	
Hoxc5	S STAATTA A			
Hoxc6	TAATTA	Antp (fly)	TRANSFAC 13 compiled seq.	AATA
Hoxc8		Antp (fly)	TRANSFAC 13 compiled seq.	AATA
Hoxc9	<u>agt</u> cATTAA	Antp (fly)	TRANSFAC 13 compiled seq.	AATA
Hoxc10	GTCSTAAA	Abd-A (fly)	TRANSFAC 40 compiled seq.	AT
Hoxc11	GTCSTAAA	Abd-B (fly)	TRANSFAC 45 Selex seq.	
Hoxc12		Abd-B (fly)	TRANSFAC 45 Selex seq.	
Hoxc13	CTCSTAAA	Cad (fly)	TRANSFAC 13 compiled seq.	
Hoxd1	TAATTA			
Hoxd3	TAATTA			
Hoxd8	TAATTAAT	Antp (fly)	TRANSFAC 13 compiled seq.	ATA
Hoxd10	GAATAAAA.	Abd-A (fly)	TRANSFAC 40 compiled seq.	A

Hoxd11		Abd-B (fly)	TRANSFAC 45 Selex seq.	
Hoxd12		Abd-B (fly)	TRANSFAC 45 Selex seq.	
Hoxd13	-FAATAAAA	Abd-B (fly)	TRANSFAC 45 Selex seq.	Geeateaa -
lpf1	TAATTA	Pdx1 (h)	JASPAR 31 Selex seq.	
lrx2	TACATOTA			
lrx3	TACATCTA			
lrx4	TACATGTA			
lrx5	TACATOTA			
lrx6	TACATGTA T			
Isl2	R			
lsx				
Lbx2	T TIMITA			
Lhx1	TTAATTAA			
Lhx2	AFTAATTA			
Lhx3	TAATTAA	Lhx3 (h)	JASPAR 21 Selex seq.	A-TAATIA-
Lhx4	S STAATTAS			
Lhx5				
Lhx6	ATTAATIA			
Lhx8				
Lhx9	TAATTAAT			
Lmx1a	TAATTAA			
Lmx1b	TAATTAA			

Meis1	A ACTOTCA	Meis1 (m)	TRANSFAC 32 Selex seq.	cTGTCA
Meox1	TAATTA			
Mrg1	ACTOTCA			
Mrg2	AGCTGTCA			
Msx1		Msx1 (m)	TRANSFAC 13 Selex seq.	
Msx2	ACCAATTA			
Msx3	- AASTAATTAA			
Nkx1-1	TTAA			
Nkx1-2	AATTA			
Nkx2-2	CACTICA	Nkx2-2 (m)	TRANSFAC 23 Selex seq.	AA=FACTTAA
Nkx2-3	- T-AAGTACTT			
Nkx2-4	CCACTA			
Nkx2-5	CCACTT_A	Nkx2-5 (m)	JASPAR 17 Selex seq.	
Nkx2-6	A CCACTTAA			
Nkx2-9	AAGTACTE A			
Nkx3-1	AAGTACTT	Nkx3-1 (h)	JASPAR 20 Selex seq.	
Nkx6-1	ATTAATTA	Nkx6-1 (r)	TRANSFAC 17 Selex seq.	
Nkx6-3				
Obox1	CONCERCISE OF CO	Abd-B (fly)	TRANSFAC 45 Selex seq.	
Obox2	GGGGATTA	Abd-B (fly)	TRANSFAC 45 Selex seq.	
Obox3	GGATTA			
Obox5				

Obox6	GGATTA			
Og2x		Nobox (m)	JASPAR 38 Selex seq.	ATTA
Otp	TTAATTAA			
Otx1	GGATTAAT	Otx1 (m)	TRANSFAC 10 compiled seq.	
Otx2	GGATTAA	Otx1 (m)	TRANSFAC 10 compiled seq.	Gattat
Pax4	z çtattaz -	Pax4 (m)	TRANSFAC 20 Selex seq.	ARTAATI
Pax6	TAATTA	Pax6 (m)	TRANSFAC 47 Selex seq.	_ I <mark>⇔QC⇔I⊴ I</mark> ≣
Pax7	- FILANIE	Prd (fly)	TRANSFAC 9 compiled seq.	AAATIR
Pbx1	CATCAAT	Pbx1 (h)	JASPAR 18 Selex seq.	ATCAATCAA
Phox2a	ATTAATTA			
Phox2b	ANTTAATTAA			
Pitx1	GGATTAA			
Pitx2		Pitx2 (h,m)	TRANSFAC 9 compiled seq.	
Pitx3	GGATTA_c			
Pknox1	ACTOTCA			
Pknox2	CTGTCA			
Pou1f1	TAATTAATTA	Pou1f1 (h,m)	TRANSFAC 17 compiled seq.	TTAT_CAT
Pou2f1	TAATTA	Pou2f1 (h)	TRANSFAC 56 Selex seq.	TATGC AAT
Pou2f2	TATCCAAAT			
Pou2f3				
Pou3f1	TAATTAATTAA			
Pou3f2	TAATTA	Pou3f2 (h)	TRANSFAC 7 Selex seq.	ITATGETAAT

Dou2f2				
Pousis	A TATCCATAA. A			
Pou3f4	AATTAATTA			
Pou4f3	ATTAATA	Unc-86 (worm)	TRANSFAC 5 compiled seq.	CATTERAAT
Pou6f1		Pou6f1 (m)	TRANSFAC 16 Selex seq.	TATTAATAS
Prop1	TAATTAA			
Prrx1	TAATTAST			
Prrx2	·	Prrx2 (m)	TRANSFAC 59 Selex seq.	TAATI
Rax	FINTAR P			
Rhox11				
Rhox6	TAATTA			
Shox2	TAATTA			
Six1				
Six2	GGTATCA T			
Six3	AT COGTATCA T			
Six4	GGTGTCA			
Six6	GGTATCA			
Tcf1		Tcf1 (h,m)	TRANSFAC 26 compiled seq.	ATTAAC
Tcf2	GITAACIA			
Tgif1	ACTOTCA -	Tgif1 (h)	TRANSFAC 15 Selex seq.	
Tgif2	A CTGTCA			
Titf1	CCACTI A	Titf1 (h,m)	TRANSFAC 7 compiled seq.	A-ACTIGA-I
Tlx2	ANT ANTTANTA	Tlx2 (h)	TRANSFAC 40 Selex seq.	

Uncx4.1	ATTAATTAA		
Vax1	TAATTA		
Vax2	ATTAATTA		
Vsx1	TAATTAA		

6. Evidence that binding profiles are independent

We evaluated whether any two 8-mer profiles were the same or different using two basic criteria. First, we considered the degree of overlap among the top-scoring 100 8-mers (determined using the E-value), as shown in Figure 2B. Two homeodomains that had an overlap within the distribution of that for duplicated experiments (see below), as well as a similar overlap with other homeodomains, were considered to be inseparable by this criterion. Second, for groups of homeodomains that were not distinct by this criterion, we considered whether differences in binding profiles across the composite set of 8-mers with E > 0.45 for any member of the group could be attributed to (a) differences in the primary motif obtained; (b) differences in the "TFspecific" motif obtained as described above, or (c) an otherwise clear theme among the 8-mers preferred or not preferred by one or more homeodomains. If at least one of these criteria supported the dendrogram obtained by clustering analysis of the 8-mer profiles, we considered it to be evidence for separable binding activities. The decisions made are outlined in the following series of figures, for each group of 8-mers. The individual class assignments are given in the Supplementary document "Homeodomain subclass assignments" in the Supplementary data. Although the assignment process here is *ad hoc*, the process is validated by the fact that the resulting groups of indistinguishable binding activities closely follow both the overall sequence similarity among the homeodomains (shown in the figures on the following pages) and the 15AA-defined groups as described in the main text, which are also listed in the document "Homeodomain subclass assignments", together with the 65-class merged PBM/15AA classifications described in the main text.

Table S2. The Top 100 overlaps among duplicated experiments were as follows:

Different gene, same homeodomain (identical at all 57 positions within the pfam-defined homeodomain but with different flanking residues):

Hoxa5-Hoxb5	86	
Lhx2-Lhx9	90	
Lmx1a-Lmx1b	85	
Vax1-Vax2	90	
lrx2-lrx5	76	
Lhx1-Lhx5	96	
Phox2a-Phox2b	91	
Pitx2-Pitx3	83	
Gbx1-Gbx2	88	
Evx1-Evx2	83	
Nkx2-4-Titf1	84	
Same gene, differen	t clones (iden	tical at all 57 positions within the pfam-defined
homeodomain but w	vith different a	mounts of flanking sequence included):
Cart1-Cart1	90	
Hoxa7-Hoxa7	76	
lrx3-lrx3	81	
Lhx6-Lhx6	86	
Obox5-Obox5	95	
Pou6f1-Pou6f1	94	
Rhox11-Rhox11	86	
Same clone, differen	nt batch of pro	otein:
Nkx6-1-Nkx6-1	91	
Six6-Six6	66	
Cutl1-Cutl1	73	
Mean		85.2
Standard Deviation		7.6
Left 0.5% of normal d	listribution	65.7

Note that proteins in these duplicate experiments were occasionally expressed by separate protocols (*in vitro* transcription/translation vs. *E. coli*). Clone sequences can be found in our supplementary table "Protein and DNA sequence", where other experimental information and the groupings derived below are also found. Expression methods for all samples are listed on <u>http://the_brain.bwh.harvard.edu/pbms/webworks2/</u>

and at

https://hugheslab.ccbr.utoronto.ca/twiki/bin/view/MouseHomeodomain/WebHome.

Figure S5. From the diagram in Figure 2B, we identified 18 groups requiring further investigation were identified (numbered A through R in the graph below, which is identical to that in Figure 2B; the color scale is the same as shown in Figure 2B, i.e. the units refer to Top 100 overlap):



Evidence for subclassification of Group A is shown in Figure 2C.

In the figures on the following pages, the dendrogram for the full homeodomain amino acid sequence is shown, but it was not relied upon in the classification procedure. A "legend" for each of the diagrams is found on the first page, followed by two pages showing the full ClustalW diagram and "dominant motif" logos.

Figure S6. This page and the next show the ClustalW phylogram tree for the homeodomain sequences for the 168 proteins for which binding specificity was determined. The dominant motif for each is also shown.





Figure S7.

The figures on the following pages contain the following for each of the groups with similar top 100 8-mer profiles (larger groups are spread over more than one page):

Group name and number of 8mers in the analysis – selected to have at least one instance of E > 0.45 among the experiments in the group

Heat-map (i.e. clustergram) with dendrograms showing hierarchical agglomerative clustering results, using E-scores for all 8-mers that have, and with Pearson correlation as the distance metric Portion of the ClustalW dendrogram, derived from the full homeodomain amino acid sequence (in order to illustrate the relationships among the protein sequences), together with the dominant motif (these are taken from the full ClustalW tree shown below)

Pull-outs showing individual 8-mers in order to illustrate sequence elements in the clusters that distinguish one or more sub-categories of proteins within the clustergram. Sequence themes here were identified manually.

Highest TF-specific 8mer E-score for each protein

TF-specific motifs

Assessment of the results shown





Group C: 187 8mers









Assessment: There are no convincing systematic differences in the 8mer sequences here, despite the three experiments on the right having a different appearance in the dendrogram.

Hoxa6	TAATTA	0.495
Hoxa7_rep1	C TATTATA	0.475
Hoxa7_rep2		0.381
Hoxdo 6	T-ASSAC ASS-	0.361
Hoxdo7		0.417
Hoxb8		0.434
Hoxc6	222 2 2 22	0.424
Hoxc8		0.467
Hoxd8		0.471

Group F: 321 8mers 0.5 0.45 0.4 0.35 Hoxa3 Hoxb3 Hoxd3 Hoxa2 Ev X1 Ev X2 Hoxa4 Hoxa5 Hoxb5 Hoxb4 Gsh2 Meox1 Emx2 Ē Hoxc4 Hoxc5 Vax1 Vax2 Colors 0.495 0.478 0.483 0.450 0.490 0.482 0.490 0.476 0.427 0.403 0.448 0,484 0.493 0.447 0.403 0.497 0.451 497 indicate groupings Ö (see next two pages) TAACGAS 00 0000 TAAzea **FAAT**GGT₁ CTAALTAG ATAATGAT TAAACGGI **CTAAcGAss** ГААТА **cTAATTAC** CGGA_AAA₃ ■TAACGAs TCATCA 4 **ATO** 1



Group F: 321 8mers with E > 0.45



Data on previous pages support the following groups: Hoxa2 Hox3 Ipf1 Evx1/2 Meox1 Hox4/5 Emx2/Vax1/2 Gsh2 Most of the TF-specific motifs agree with the 8mer clustering. Some of the distinguishing features in the clustergrams are *unpreferred* sequences.



















Group M: 288 8mers





The Obox proteins seem to require more G's.

Try repeating the analysis without the Obox proteins, which are obscuring subtler differences between the other proteins. Over ->

Group M2: 255 8mers



TF-specific motifs are shown in the next page ->

	_ •	
Crx		0.492
Dmbx1		0.490
Otx1		0.490
Otx2	A TAAGee	0.480
Gsc		0.497
Dobox5	S STAAGCC	0.494
Pitx1		0.497
Pitx2	TAAGec =_	0.490
Pitx3	CTAAGCC	0.497
Obox1		0.484
Obox2		0.431
Obox3		0.365
Obox5_rep1	$\hat{\tau} = - \mathbf{T} \mathbf{S} \Delta \hat{\mathbf{C}} \hat{\mathbf{C}}$	0.384
Obox5_rep2	T_&&I&I&II	0.397
Obox6		0.491

Evidence here indicates four groups:

Gsc

Obox1/2/3/5/6 Pitx1/2/3/Dobox5 Otx1/2/Crx/Dmbx1 (Dmbx1 is a candidate for its own group)









Group P: 180 8mers





These are the TF-
specific motifs for
Group A, shown in
Figure 2D

Ab3	I JEATTA	0.469
Alx4		0.463
Phox2a	SE TAATTCGC	0.462
Phox2b	TTAATTAA T	0.474
Pmx2		0.484
Arx		0.483
Uncx4.1	AATTAATTAA	0.494
Shox2		0.466
Esx1		0.492
Prrx1		0.477
Cart1		0.451
Otp	- TTAATTAA	0.499
Rax		0.485
Vsx1	TIGATTAA	0.485
lsx	- CIAAITGeese	0.483
Lhx1	TTAATIAAT_	0.492
Lhx5	TAA TAAA	0.497
Lhx3	TTAATTAA	0.497
Lhx4		0.492
Lhx2	_ TAAçGAGe	0.496
Lhx9		0.492
Nkx1-1	CTAATege	0.495
Nkx1-2	CTATTAG-I	0.484
Msx1		0.464
Msx2	_ <u>stAATTGete_</u>	0.464
Msx3		0.498
Gbx1	CTAATTGCTET	0.494
Gbx2	TAAICGT	0.415
Hbb9	CTAAICG_TT	0.493
Hoxd1		0.496
Pax7		0.494
En1	A_GATACG	0.434
En2	CTUAAICA	0.496
Dbx1		0.481
Dbx2		0.485
Dbx3		0.487
Dbx4		0.478
Dbx5	TAATIGC	0.496
Prop1		0.463
Lbx2	TAAC_AG_	0.497
Og2x	CAATIGA	0.491
Hoxa1		0.439

7. Predicting 8-mer profiles and scoring the predictions

We considered two general methods for predicting 8-mer binding profiles on the basis of the primary amino acid sequence: nearest-neighbor and linear regression. These both have the advantages of being able to make quantitative predictions from categorical features, and fast run times.

Nearest Neighbor

In the nearest neighbor (NN) approach, we predict the 8-mer profile of any given homeodomain protein by taking the 8-mer profile(s) of its nearest neighbor(s) (averaging E-scores in the case of a tie). We tried several variations, which differed from each other in the distance metric used and/or the residues considered to determine the nearest neighbor relationship. The two distance metrics used were: (a) the number of different AAs between two aligned homeodomains, and (b) the sum of the negative of the similarities (as given by the PAM250 matrix) between the AAs of two aligned homeodomains. With respect to the residues considered, we tried a "full-homeodomain" version where all 57 residues in the pfam-defined homeodomain model were used to determine the nearest neighbor(s) of a protein and several "specificity residues" versions where four different sets of residues known in the literature (10, 11) to make direct contact with the DNA where used to determine the nearest neighbor(s) of a protein (see below).

Linear Regression

Linear regression techniques require a vector of numerical values as input, and the number of features should be at least five-fold smaller than the number of examples. In our case, the features must represent the protein sequence, and should preferably number less than ~30, since we analyzed 168 homeodomain proteins. The full homeodomain family protein sequence alignment was downloaded from Pfam (Pfam Accession Number: PF00046), from which we extracted only the alignments of the 168 mouse homeodomains. This alignment was then converted to a binary representation by replacing all 20 standard amino acids in any of the canonical residue positions with unique 20-bit binary flags. Regions of the Pfam alignment that correspond to insertions were treated as separate binary variables, with a value of 1 indicating the presence of an insertion in that particular region. We then applied Principal Components Analysis (PCA) (12) to the alignment encoding in order to reduce the number of variables per protein in this encoding strategy, and to eliminate correlation between these variables. The number of components retained (23) was selected using Parallel Analysis (PA), which is essentially a random permutations test that asks whether the Nth component explains more of the variance than the Nth component would in a permuted version of the same data (reviewed in reference (13)). These 23 components together account for 70% of the variance in the binary vectors.

For a given 8-mer, we used Partial Least Squares (PLS) regression (14) to model the relationship between the retained principal components and the 8-mer Z-score, which are the independent variables and the response variable, respectively (we used Z-scores for this analysis, under the assumption that the model would learn a linear relationship between sequence features and affinity, which, as described above, is more likely to be reflected by Z-scores than E-scores). An internal round of cross-validation (15) was used to determine both the optimal model and the optimal number of latent variables (see reference (16) for details of PLS). Finally, for each of the 168 proteins, we predicted its Z-score profile across all 8-mers using a leave-one-out crossvalidation strategy, wherein a distinct model learned for each 8-mer and for each homeodomain (i.e. a separate model for all 157 x 32,896 entries in the data table) was used to predict its Z-score entry.

We also tested other regression approaches (Support Vector Regression, Lasso, and Ridge Regression) and other encodings of the amino acid sequences; the PLS results are shown because they were the highest-performing. A full description of our efforts to use regression techniques to model PBM data will be published elsewhere.

Comparison of prediction methods

Each NN version and the regression outputs were evaluated by leave-one-out cross-validation using the median and the mean of the following performance measures: the Spearman correlation coefficient over all 32,896 E-scores (Z-scores in the case of linear regression), the number of top-100 8-mers in common, and the Root Mean Square Error (RMSE) between the predicted and the measured profiles (E-scores for all methods except linear regression which used Z-scores). The 1st and 2nd place in each scoring category (rows) are indicated by dark and light green, respectively. In the table below, "A vs. A" indicates that the evaluations were done between the actual real and predicted homeodomain (for example, comparing the predicted Hoxa13 profile to the real Hoxa13 profile). "All vs. all" indicates all possible pairings among all homeodomains in the data set, i.e. the aggregate of all predicted A vs. real B homeodomains (for example, comparing the predicted Pdx1 8-mer profile to the actual Hoxa13 8-mer profile). This statistic is used as a negative control to ensure that the models are not simply learning an average 8-mer profile over all homeodomains, which would result in (low) positive values for all measures because on average there is a positive correlation between randomly-selected experiments. We also considered the difference between the "A vs. A" statistics and the "All vs. all" statistics as a measure of specificity ("Difference between A vs. all medians" in the table). Wins were tallied as 1 for first-place (including ties) and 0.5 for second place (allowing ties). We note that the high 15AA-selected NN win tally is robust to most variants of our scoring scheme that include the Top 100 criterion.

Approach		NN	Residue- selected NN	Residue- selected NN	Residue- selected NN	Residue- selected NN	Residue- selected NN	Residue- selected NN	Residue- selected NN	Residue- selected NN	Partial Least Squares
Residues		All	Set 1	Set 2	Sets 2,3	Sets 2,3,4	Set 1	Set 2	Sets 2,3	Sets 2,3,4	NA
Similarity metric		match = 1, mismatch = 0	match = 1, mismatch = 0	PAM250	PAM250	PAM250	PAM250	NA			
Spearman	median	0.83	0.88	0.86	0.86	0.84	0.88	0.86	0.87	0.85	0.85
real A	mean	0.8	0.85	0.83	0.84	0.82	0.85	0.83	0.84	0.82	0.83
Spearman all	median	0.65	0.72	0.74	0.69	0.68	0.72	0.74	0.7	0.68	0.74
real (control)	mean	0.63	0.69	0.7	0.66	0.65	0.69	0.7	0.66	0.65	0.7
Spearman difference A vs. all medians		69	65	36	72	72	67	40	72	69	52
Top100-overlap	median	81	77	69	82	82	76	70	82	82	74
real A	mean	72.9	69.1	62.4	75.2	75	68	62.3	74.5	74.7	67.2
Top100-overlap	median	12	12	33	10	10	9	30	10	13	22
all predicted vs all real (control)	mean	26.7	28.4	31.6	26.5	26.3	27.6	31.3	26.7	27.1	28.5
Top100-overlap difference A vs. all median		69	65	36	72	72	67	40	72	69	52
RMSE predicted	median	0.65	0.57	0.63	0.56	0.62	0.57	0.63	0.57	0.61	0.62
A vs real A	mean	0.76	0.63	0.68	0.63	0.66	0.63	0.68	0.64	0.67	0.66
DMCE all	median	1.17	1.02	0.97	1.07	1.09	1.01	0.96	1.06	1.08	0.96
RMSE all predicted vs all real (control)	mean	1.22	1.06	1.03	1.12	1.14	1.05	1.01	1.11	1.13	1.03
	range	[0.0-3.19]	[0.0-2.82]	[0.0-2.88]	[0.0-3.19]	[0.0-3.19]	[0.0-2.82]	[0.0-2.87]	[0.0-3.19]	[0.0-3.19]	[0.31-2.78]
RMSE difference A vs. all median		-0.52	-0.45	-0.34	-0.51	-0.47	-0.44	-0.33	-0.49	-0.47	-0.34
No. of proteins with an overlap (predicted A vs real A) < 50		17	31	41	14	15	31	41	15	16	28
Wins		5	4	0	9	7	5	0	7	3.5	0

Table S3. Scoring metrics for predicting 8-mer profiles.

Residues considered were as follows:

1) 47, 50, 54 (traditional specificity residues)

2) 3, 5, 47, 50, 51 (major or minor groove contacts from Engrailed structure (10))

3) 6, 25, 31, 44, 46, 48, 53, 54, 55, 57 (phosphate backbone contacts from Engrailed structure (10))

4) 7, 8, 28, 43, 52 (positions that contact DNA in other homeodomain structures)

8. Consistency between homeodomain groups derived from PBM data and homeodomain amino acid sequences.

Our initial grouping of homeodomains on the basis of 8-mer profiles, in which we first compared the top 100 8-mers and then investigated 19 groups for systematic differences (i.e. TF-specific motifs, or clear differences in the primary motif) among all binding 8-mers within the group, resulted in definition of 71 different groups (all group IDs are found in the supplementary document "Homeodomain subclass assignments") among which 31 have only one member. Our initial grouping on the basis of identity among the 15 selected amino acid positions (allowing no mismatches) resulted in 74 different groups, among which 40 have only one member.

The correspondence between these initial groups was as follows:

- Of the 71 different groups defined on the basis of 8-mer preferences, 59 of them consisted entirely of proteins that are identical among all 15 amino acids, i.e. they are in the same initial amino acid-based group.
- Of the 74 different initial amino acid groups, i.e. sets of proteins with the same residues at all 15 DNA-contacting residues, 50 of them consisted entirely of proteins that are in the same category as defined by the 8-mer profile.
- There are 42 groups that are completely consistent between the two categorizations. 27 of these groups consist of a single protein.

We reasoned that disagreement between the two categorizations might be due to the divisions between categories being more stringent in one classification scheme relative to the other. This would be unavoidable (a) if more than one configuration of the 15 amino acids could result in the same 8-mer binding profile, or (b) if additional amino acids besides the 15 selected influence the 8-mer binding profile.

We therefore asked whether simply merging groups in one categorization or the other could result in a more uniform set of categories, by grouping homeodomains that have either 8-mer profiles that are more highly correlated with the group they are merging into then they are to any other homeodomain in the data set, or by grouping those with the most closely related (but not identical) set of 15 amino acids (the groupings above allowed no mismatches among the 15 amino acids, and therefore reflect only identity, and not similarity). Indeed, by allowing the following set of merges, and one reassignment (Hoxc12, which is an unusual case) we obtained 65 groups that are entirely consistent between the fifteen amino acid residues and the 8-mer binding profiles.

1. Merge Poulflwith Pou3f1, Pou3f2, and Pou3f4. Poulf1 is one amino acid different from the others among the 15AA but has an indistinguishable DNA-binding activity.

- 2. Group Hoxc12 with the Hox9, 10, and 11 group, instead of with Hoxd12. Hoxc12 is more similar to Hoxd12 over the entire homeodomain, but it is identical to the Hox9,10,11 group on the basis of the 15 amino acids, whereas it has two amino acid differences among the 15 DNA-contacting residues. The data figure in Section 5 suggests that its 8-mer binding profile has characteristics of the Hox9,10,11 group, as does its TF-specific motif.
- 3. Group Hoxb13 with the rest of the Hox13 group. It contains an amino acid mismatch among the 15 selected residues, but the 8-mer profile is indistinguishable by our criteria.
- 4. Group Esx1, Isx, and Otp with Alx4, Arx, Cart1, Phox2a, Phox2b, Prrx1, Prrx2, Rax, Shox2, and Uncx4.1. The 8-mer profile is indistinguishable by our criteria and these three homeodomains have only one amino acid mismatch from the rest of this group among the 15 selected residues.
- 5. Group Prop1 with Alx4, Arx, Cart1, Phox2a, Phox2b, Prrx1, Prrx2, Rax, Shox2, and Uncx4.1 group. Its 8-mer profile appears to be distinct, indicating that additional amino acids influence binding specificity, but it is indistinguishable by the Top 100 overlap and has the same 15 DNA-contacting residues as the remainder of this group.
- 6. Group Obox6 with Obox1, Obox2, Obox3, and Obox5. It has three amino acid differences among the 15 DNA-contacting residues, but the 8-mer profile is indistinguishable by our criteria.
- 7. Merge all members of the Six class (Six1, Six2, Six3, Six4, and Six6). The 8-mer profiles are indistinguishable by our criteria, and they all share 12 common amino acids among the 15 DNA-contacting residues.
- 8. Merge all members of the Irx class (Irx2, Irx3, Irx4, Irx5, Irx6). The 8-mer profiles are indistinguishable, and they all share 12 common amino acids among the 15 DNA-contacting residues.
- 9. Merge Pknox1 and Pknox2. The DNA-binding profiles were called as distinct because they correlate more highly with other experiments than they do with each other, but do they appear in the figure in Section 5 to have similar activities, and they have no amino acid differences among the 15 DNA-contacting residues.
- 10. Merge Hlxb9, and all members of the Hox3, 4, and 5 families. These clearly form three distinct groups on the basis of the 8-mer profile, but there is substantial overlap among the top 100 8-mers, and they are all identical at the 15 selected amino acid residues. This is a group in which the 15 amino acids clearly do not completely govern the entire DNA binding specificity.
- 11. Merge Hoxa2 and Ipf1. These have only a single amino acid difference among the 15 DNA-contacting residues (Valine vs. Isoleucine at P47) and their 8-mer binding profiles are indistinguishable.
- 12. Group Pou3f3 with Pou2f1, Pou2f2, and Pou2f3. There are no differences among these four at the 15 DNA-contacting residues.
- 13. Group all members of the Hox9, 10, and 11 families. These clearly form three distinct groups on the basis of the 8-mer profile, but there is substantial overlap among the top 100, and they are all identical at the 15 selected amino acid residues. This is a group in which the 15 amino acids clearly do not completely govern the entire DNA binding specificity.

- 14. Group Dbx1 and Dbx2. These appear distinct on the basis of their 8-mer binding profiles, but there is substantial overlap among the Top-100 8-mers (62) and they are identical at the 15 DNA-contacting residues.
- 15. Group En1 and En2 together with Gbx1 and Gbx2. These two pairs appear distinct on the basis of their 8-mer binding profiles, but there is substantial overlap among the Top100 (they are identical by the top 100 criterion) and they are identical at the 15 DNA-contacting residues.
- 16. Group Lmx1a and Lmx1b together with Lhx3 and Lhx4. These two pairs appear distinct on the basis of their 8-mer binding profiles, but they are identical at the 15 DNAcontacting residues. This is a group in which the 15 amino acids clearly do not completely govern the entire DNA binding specificity.
- 17. Group all members of the Nkx2 class (including Titf1). There are clearly two sub-classes on the basis of the 8-mer binding profiles, but they are all identical at the 15 DNA-contacting residues. This is a group in which the 15 amino acids clearly do not completely govern the entire DNA binding specificity.

The results of these operations, including tentative names for each of these subclasses, are included in the document entitled "Homeodomain subclass assignments".

9. Agreement between Predicted Z-score vs. measured relative affinity for the *Drosophila* Engrailed homeodomain

Figure S8. The scatter plot shows the predicted Z-score on the X-axis, and the relative affinity from Damante *et al.* (*17*) on the Y-axis. Z-scores (which we take to be our most accurate inference of binding affinity) were inferred following the same protocol as for E-scores (nearest-neighbor over the 15 DNA-contacting residues, with ties averaged) with the Z-score for each 9mer taken as the average of the Z-score for the two overlapping 8-mer components.



10. Comparison between PBM data and ChIP-chip or ChIP-seq data

We analyzed six ChIP-chip or ChIP-seq datasets for homeodomain proteins available in the literature (see Figure 7 in the main paper and the Table and Figure on the next two pages). Bound sequences were scanned to determine enrichment of highly-bound PBM 8-mers. To standardize the length of the bound sequences across all datasets, we took either the ChIP peak (if known) or the center of the identified bound sequence and add 1 kb on each side. Enrichment ratio was determined with respect to 2kb-length random genomic regions taken from the genome (same version as the one used for the chip experiments) of the corresponding organism. For each dataset, the number of random genomic regions sampled was 10 times the number of bound sequences.

In the plots on the following page, random and bound sequences were scanned with the predicted PBM E-score 8-mer profile (since none of these data are from mouse) using a 500 bp length moving window with a 50 bp tiling distance. For each dataset, the E-score with the highest enrichment ratio in the central portion were selected by sampling 20 intervals starting from E = 0.43 to the maximum value. Enrichment ratio was calculated as the ratio of number of 8-mers above the cutoff value found in the bound sequences to one tenth of the number of 8mers above the cutoff value found in the random genomic sequences.

Homeodom ain	ChIP data reference	Organism ChIP experimen ts	# of bound sequences	Is ChIP peak identified in original reference?	8-mer profile	# mismat ches with NNs	E- score thresh old
Tcfl/Hnfla	Odom et al, MSB 2006 (PMID: 16738562)	Human	427 *	No	Predicted	1	0.456
Caudal/Cdx2	Li et al, PloS Biol. 2008 (PMID: 18271625)	Fly	1331	Yes	Predicted	0	0.493
Pou5f1/Oct4	Boyer et al, Cell 2005 (PMID: 16153702)	Human	603	No	Predicted	1	0.487
Nanog	Boyer et al, Cell 2005 (PMID: 16153702)	Human	1554	No	Predicted	1	0.477
Bagpipe/Bap x1	Jakobsen et al, Genes Dev. 2007 (PMID: 17908931)	Fly	78	No	Predicted	0	0.491
Barx2	Stevens et al, J Biol. Chem 2004 (PMID: 14744868)	Human	42	No	Measured	0	0.489

Table S4. Chip-chip data sets analysed.

* Bound sequences not provided by Odom *et al*. Bound sequences were determined by analyzing their raw microarray data with an in-house software similar to the one described in Boyer *et al*.

Figure S9. For the data sets in Table 4, random and bound sequences were scanned with the predicted PBM E-score 8-mer profile using a 500 bp length moving window with a 50 bp tiling distance.



11. Supplemental References

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