

Supplementary Information for

Integrated functional genomic analyses of Klinefelter and Turner syndromes reveal global network effects of altered X chromosome dosage

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Supplementary Information Text

Samples

Individuals with SCAs were recruited through national organizations, including the Turner Syndrome Society and Association for X and Y Variations; a local network of physicians; and advertisement on the Stanford University School of Medicine website. Typically developing control participants were recruited locally. Exclusion criteria for both groups included premature birth (gestational age < 34 weeks), low-birth weight (< 2000 g), and known diagnosis of a major psychiatric or neurological condition. Participants in SCA cohorts were recruited after review of clinical records provided by the family to confirm karyotypes. Mosaic karyotypes and structural abnormalities were excluded, only subjects with monosomic karyotypes as determined by standard karyotype analysis of at least 20 cells (1). Participants were part of a larger study at Stanford University examining neuroanatomical development in TS and KS; peripheral whole blood was obtained from a subset of participants. Fifty-five individuals — 14 females with Turner Syndrome (mean age \pm SD: 8.27 \pm 1.83), 14 males with Klinefelter syndrome (mean age \pm SD: 7.50 \pm 2.21), 13 age-matched male (mean age \pm SD: 6.93 \pm 2.27) and 14 age-matched female (mean age \pm SD: 7.71 ± 1.77) typically developing controls — were included in the present study.

Peripheral blood mononuclear cells (PBMCs) were isolated from whole blood samples by centrifugation after dilution with PBS and underlaying Sigma Histopaque-1077 to the bottom. Genomic DNA and total RNA were extracted simultaneously from PBMCs using QIAGEN's AllPrep DNA/RNA Mini Kit.

Genotyping

DNA samples were genotyped on the Infinium Multi-Ethnic Global-8 arrays (Illumina, Inc., San Diego, CA, USA). Raw probe intensity data were normalized and processed by Illumina's GenomeStudio software (version 1.9.4) to export the genotype, the log R ratio (LRR) and B allele frequency (BAF) for all markers for each sample. Genotyping coordinates were based on hg38. Genotype call rate is above 99.8% for all samples. LRR and BAF were plotted against physical positions to confirm the number of X chromosome for all individuals. All TS patients but one carried one X chromosome indicated by decreased LRRs of X chromosome relative to female controls. All KS patients carried two X chromosomes indicated by increased LRRs relative to male controls. The one misdiagnosed TS patient was excluded from the study. KS patients with two distinct X chromosomes were shown by three bands of BAF plots and those have identical ones were shown by two bands of BAF (Figure S1).

RNA-Seq library preparation and sequencing

Illumina's TruSeq Stranded RNA with Ribo-Zero (Human) for ribo-depletion kit was used to construct strand specific libraries for each sample. Sequencing was performed on Illumina's HiSeq2500 system with 100 bp paired-end sequencing method. An average of 37M paired-end reads was generated for each sample.

RNA-Seq data analysis

Cutadapt (2) (version 1.8.1) was used to trim Illumina TruSeq adapters and low-quality ends from the raw reads. Bowtie 2 (3) (version 2.3.1) was used to align the trimmed reads to the GENCODE comprehensive gene annotation (release 23, hg38) and RSEM (4) (version 1.2.30) was used to quantify gene expression in a strand-specific manner by setting parameter "--forward-prob 0". limma (5) (version 3.33.7) was used to perform differential expression analysis. Genes with FDR adjusted *p*-value < 0.1 were considered to be significant. Only the genes with detectable expression (FPKM > 1) were used for differential expression analysis, WGCNA and pathway enrichment analysis. The X-inactivation/escape status of the X chromosome genes were obtained from a previous study (6).

Global expression of X chromosome and autosomes for each individual were calculated by medians of the FPKM values of the remaining genes after applying different minimum expression levels. Genes in PARs were excluded from this analysis as they are dosage-sensitive. The ratio of X chromosome expression to autosomal expression (X:A ratio) was calculated for each individual separately.

Allele-specific expression analysis

Samtools (7) mpileup (version 0.1.19) and BCFtools (version 0.1.19) were used to count the number of reads mapped to each allele of the heterozygous single nucleotide variants (SNVs) genotyped by Infinium Multi-Ethnic Global-8 arrays. Binomial tests were performed to test if the percentage of RNA-Seq reads mapped to the alternative allele is significantly different from the mean frequency of alternative allele of all heterozygous SNVs within each sample. Only the SNVs with coverage > 10 in RNA-Seq data were included in the analysis. Fisher's exact test was performed to identify the expressed genes that have different patterns of allele-specific expression between patients and their controls.

Expression quantitative trait loci (eQTLs) analysis

Association testing for eQTLs was conducted using the R package Matrix eQTL (8) (version 2.1.1). Only SNVs with minor allele frequency (MAF) > 0.1 from 1 Mb upstream to 1 Mb downstream of a gene were analysed. Both additive linear model and interaction model were tested.

Functional annotation and co-expression network analysis

DAVID (9) (version 6.7) was used to conduct pathway enrichment analysis using all the genes with detectable expression as background to identify the biological processes that were significantly enriched with differentially expressed genes in each comparison. Biological processes with bonferroni corrected p values smaller than 0.05 were considered significant.

To identify modules of co-expressed genes, we performed weighted-gene coexpression network analysis (WGCNA, version 1.46) (10) using all the genes with detectable expression (approximately 14,000 genes). Normalized count data obtained by the "varianceStabilizingTransformation" function in DESeq2 (11) were used. Signed network was constructed for each comparison.

TF binding site enrichment analysis

We first collected TFs with known motifs from existing databases and previous publications including Homer (12), Jaspar (13), ENCODE (14) and Jolma et al. 2013 (15). Then we identified target genes of these TFs and regulatory elements (*e.g.* enhancers) through transcriptional regulatory network in primary B cell and T cell from PECA model (16), which is an efficient statistical approach to infer transcriptional regulatory network based on paired expression and chromatin accessibility data across diverse cellular contexts. Motif enrichment on regulatory elements of the DEGs (upregulated and downregulated DEGs separately) was performed using Homer in both TS and KS.

Captured bisulfite sequencing libraries

Library preparation was performed using Roche's SeqCap Epi CpGiant Probes kit (the size of target regions is 80.5 Mb with > 5.5 million CpGs) as previously described (17). Briefly, 1μg of genomic DNA was sheered to fragments of 180-220 bp by sonication and was used to construct library with the KAPA library Preparation Kit Illumina platforms (KAPA biosystems). DNA libraries were bisulfite converted, purified and amplified, followed by hybridization with SeqCap Epi probes.

Methylation data analysis

Captured bisulfite converted libraries were sequenced on Illumina's NextSeq 500 platform by 2x150 paired-end sequencing with an average of 88 million reads generated for each sample. After trimming the adapters and low-quality ends by Cutadapt, the reads were mapped to human RefSeq genome (GRCh38.p10) using Bismark (18) (version 0.16.3). Duplicates were removed by the deduplicate bismark script in Bismark. Only one copy of the overlapping parts in the middle of paired-end reads was retained after clipping the read with the lower average quality in the overlap region by the "clipOverlap" tool in bamUtil (version 1.0.14; https://github.com/statgen/bamUtil). Ontarget read rate and coverage were calculated by Qualimap (19) (version 2.1). The average ontarget read rate was 63.7% and average on-target coverage was 41.2X across the samples.

Methylation ratio for each CpG was extracted by the bismark methylation extractor script in Bismark. For each sample, only CpGs with at least ten reads covering them were included in the downstream analysis. Differentially methylated regions (DMRs) were identified between any two cohorts by metilene (20) (version $0.2-6$) with $>= 3$ CpGs and a mean methylation difference between the two compared groups of >= 0.2. DMRs with FDR corrected *p*-value < 0.05 were considered significant. All the DMRs were assigned to genes whose transcription start sites (TSSs) were closest to them. Promoters of the genes were defined as regions from 1.5 kb upstream of TSSs to 0.5 kb downstream.

Analysis of in situ Hi-C maps.

Four LCLs including one TS patient (GM23916), one KS patient (GM17879), one female control (GM12878) and one male control (ID00016) were acquired from Coriell cell repository. In situ Hi-C libraries were prepared using the previous reported protocol (21) and sequenced on Illumina's HiSeq 4000 system with 150 bp paired-end sequencing method. An average of 308 million read pairs was generated for each sample.

Hi-C reads were mapped to reference genome (hg38) by BWA-MEM (22) (version 0.7.7) in single end mode after ligation junction removal using Cutadapt (2) (version 1.8.1). Uniquely mapped reads were paired up and filtered (*e.g.* duplicates, self-ligations, continuous genomic fragments or re-ligation events, regions with abnormally high concentrations in reads) using HOMER (12) (version 4.8). Raw contact maps were constructed using HOMER (12) and corrected by total number of contacts for each sample. The haploid contact map of the inactive X chromosome in KS patient was obtained by subtracting the corrected contacts of male control from the corrected diploid contacts of KS patient, while the haploid contact map of the inactive X chromosome in female control was obtained by subtracting the corrected contacts of TS patient from the corrected diploid contacts of female control. The coverage corrected contact maps were further normalized by Knight-Ruiz Matrix Balancing algorithm using gcMapExplorer (23). A/B compartment were called by HOMER (12).

Deconvolution analysis.

Cell type deconvolution of methylation sequencing data was conducted by R package EpiDISH (24), which allows us to infer the fractions of seven immune cell subtypes (B cells, NK cells, CD4 T cells, CD8 T cells, monocyte, neutrophils, eosinophils) in our samples using whole blood subtypes reference "centDHSbloodDMC.m". Deconvolution of RNA-seq data was performed by quanTIseq (25). quanTIseq can take transcripts per millions (TPM) as input and quantify via deconvolution the proportions of ten different immune cell types: B cells, Classically activated macrophages (M1), Alternatively activated macrophages (M2), Monocytes, Neutrophils, Natural killer (NK) cells, Non-regulatory CD4+ T cells, CD8+ T cells, Regulatory CD4+ T (Treg) cells, Dendritic cells.

Figure S1. Plots of Log R Ratio (LRR, upper panels) and B Allele Frequency (BAF, lower panels) for typical TS, KS patients and euploid controls. Left panels show X chromosome; right panels show chromosome 10 as an autosome control. (A) A female control. (B) A TS patient. (C) The one misdiagnosed TS patient. (D) A male control. (E) A KS patient with two distinct X chromosomes. (F) A KS patient with two identical X chromosomes. Typical TS patients carry one X chromosome indicated by decreased LRRs of X chromosome relative to female controls. Typical KS patients carry two X chromosomes indicated by increased LRRs relative to male controls. KS patients with two distinct X chromosomes are shown by three bands of BAF and those have identical ones are shown by two bands of BAF.

X0 vs. XX pathway

Figure S2

7

Figure S2. Pathway enrichment analysis of differentially expressed genes in X0 vs. XX (A) and in XXY vs. XY (B). Biological processes with bonferroni corrected *p*value < 0.05 were considered significant (44 for X0 vs. XX and 2 for XXY vs. XY). Only the uncorrected -log10 *p*-values of the top 20 biological processes are shown for X0 vs. XX.

Figure S3. Global expression of the X chromosome (A) and X/A expression ratio (B). The median of FPKM values for X chromosome genes was calculated for each individual separately. Shown in (A) are the mean and associated 95% confidence interval of the medians across all individuals within each group. The median of FPKM values for autosomal genes was also calculated for each individual. The ratio of the X chromosome to autosomal expression (X:A ratio) was calculated by (median FPKMs of X chromosome genes)/(median FPKMs of autosomal genes) for each individual separately. Shown in (B) are the mean and associated 95% confidence interval of the ratios across all individuals within each group. 9

X0 vs. XY DEGs

XXY vs. XX DEGs

Figure S4. Differential expression analysis between TS patients and male controls (A-C), and between KS patients and female controls (D-F). -log10(*p*-values) across the genome are shown in (A) for X0 vs. XY and in (D) for XXY vs. XX. Genes in PAR1 are colored in black. -log10(*p*-values) across the X chromosome are shown in (B) for X0 vs. XY and in (E) for XXY vs. XX. Genome-wide significance is based on FDR < 0.05 indicated by the horizontal lines. PAR1 region is represented by the vertical black lines and genes are shown in four colors based on their XCI status in (B) and (E). Log2 fold change and 95% confidence interval of DEGs on the X chromosome are shown in (C) for X0 vs. XY and in (F) for XXY vs. XX. DEGs are shown in four categories based on their XCI status. Genes on the Xp and the Xq are underlined by solid and dashed line respectively within each category. DEGs shared between two comparisons are highlighted in bold.

Figure S5. Shared DEGs between X0 vs. XY and XXY vs. XX. DEGs on the X chromosome are shown in four categories based on XCI status: escape, variable escape, inactive and unknown status. Dashed black circle represents log2 fold change = 0.

Pathways enriched in yellow module in X0 vs. XX

Pathways enriched in cyan module in X0 vs. XX

Figure S6. **WGCNA of X0 and XX.** (A) Module significance defined as the mean gene significance across all genes in the module. The higher the module significance, the more significantly related the module is to the TS. (B) Association of module eigengenes with the disease status. –log10 *p-*values are shown. Modules above the significance cutoff indicated by the dashed black lines are associated with TS. The grey module is reserved for genes which are not part of any co-expressed module. (C) Pathways enriched in genes of yellow module. (D) Pathways enriched in genes of cyan module.

C Pathways enriched in light-yellow module in XXY vs. XY

Figure S7. WGCNA of XXY and XY. (A) Module significance defined as the mean gene significance across all genes in the module. (B) Association of module eigengenes with the disease phenotype. –log10 *p-*values are shown. No module was significantly associated with KS after multiple testing correction. The grey module is reserved for genes which are not part of any co-expressed module. (C) Pathways enriched in genes of light-yellow module.

Figure S8. WGCNA of four cohorts together. (A) Module significance defined as the mean gene significance across all genes in the module. (B) Association of module eigengenes with the sex chromosome number $(X0:XX:XY:XXY = 1:2:2:3)$. Modules above the significance cutoff indicated by the dashed black lines are associated with sex chromosome number. The grey module is reserved for genes which are not part of any co-expressed module. (C) Normalized expressions of genes in purple module. Expressions increase with the number of sex chromosome. (D) Normalized expressions of genes in light-yellow module. (E) Normalized expressions of genes in blue module. Expressions decrease with the number of sex chromosome for genes in light-yellow and blue module.

Figure S9. The PECA subnetwork regulating inactive genes on X chromosome by differentially expressed TFs in X0. Eight inactive genes on X chromosome were regulated by *ZFX* directly, or by differentially expressed TFs on autosomes. Of which, inactive genes *FAM50A, GDI1, ARMCX3, CYBB, ARMCX6* and *MOSPD2* are expressed consistently with annotation (activation/repression) by PECA model.

Figure S10. Methylation levels of the CpGs on X chromosome. The color scale of the heatmap goes from 0 (blue) to 0.5 (white) to 1 (red). The sample names are shown on the left starting with the karyotype (X0, XX, XXY or XY).

Figure S11. Venn diagram of inactive genes (A) and escape genes (B) with **differentially methylated promoters in X0 vs. XX and XXY vs. XX.**

Figure S12. A/B compartment of X chromosome in TS and KS patient with their euploid control. Shown are the eigenvector values from PCA of correlation matrix of the contacts at 100 kbp resolution. Dashed black line represents boundary of the two superdomains in the contact map of the inactive X chromosome in female control and KS patient.

■B.cells ■Macrophages.M1 ■Macrophages.M2 ■Monocytes ■Neutrophils ■NK.cells ■T.cells.CD4 ■T.cells.CD8 ■Tregs ■Dendritic.cells

B

Figure S13. Cell type deconvolution of RNA-Seq data. (A) Relative cell type fractions of ten immune cell types inferred by quanTIseq. (B) PCA plot of all individuals based on their inferred cell compositions.

Figure S14. Cell type deconvolution of methylation sequencing data. (A) Relative cell type fractions of six immune cell types inferred by EpiDISH. (B) Student's t-test *p*values of X0 vs. XX and XXY vs. XY for each cell type.

Table S1. Log2 fold changes of shared DEGs across comparisons of **X0 vs. XX, XXY vs. XY, and XX vs. XY.**

Table S2. Distribution of DEGs in XO vs. XX on X chromosome and autosomes in WGCNA modules.

Table S3. Distribution of DEGs in XXY vs. XY on X chromosome and autosomes in WGCNA modules.

Table S4. Distribution of shared DEGs in TS and KS on X chromosome and autosomes in WGCNA modules.

Table S5. Differential expression of X-Y homolog genes between TS and female controls.

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