Dynamic-array based method for high throughout and flexible assessment of pluripotency in PSCs

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Background and Objectives: The mechanisms responsible for maintaining pluripotency during expansion of pluripotent stem cells (PSCs) to date remain unclear. Continuous propagation of PSCs is associated with varying levels of spontaneous differentiation as well as genomic aberrations. Any changes to the quality of the cells have to be detected in order to pass or fail produced batches of cells (e.g. cell banks). Differentiation of pluripotent cells can be detected by gene expression analysis, where reduction of expression of self-renewal genes and increased levels of expression of differentiation markers is observed. Current methods such as hPSC Scorecards[™], can detect changes by comparing samples to reference material, however, scorecards also require high quantity of RNA for analysis. We therefore aimed to develop a method which would allow more flexible, high throughput analysis for pluripotent gene expression analysis by applying dynamic array methodology.

Comparison of a dynamic array gene expression panel against Scorecards $^{ m TM}$ in PSCs directed differentiation assay

Methods

Design: Cells from a CGT-RCiB10 Master Cell Bank (MCB) were profiled by gene expression using hPSC Scorecard™ (Life Technology; Table 1) and by dynamic array using a CGT panel of TaqmanTM gene expression assays (Figure 1). Cell potency was assessed by differentiation into the 3-germ layers of embryoid bodies (EBs) cultured in serum-containing medium. Cells were cultured for 3 passages, then divided into 6 well plates. Non-differentiated cells were harvested on day o. Ectoderm, endoderm and mesoderm differentiated cells were harvested after 6 days of directed differentiation (Figure 1) RNA was extracted using Qiagen mini Rneasy kits, and reverse transcriptase was performed using Quantitect RT (Qiagen). RNA (8µg) was used for each scorecard sample and the equivalent of 0.5µg RNA was used for dynamic array samples. Scorecards were run on Quantstudio 7 (life technologies) and 48.48 dynamic arrays were run on Biomark HD (Fluidigm). Statistical analysis was performed using Genex 6 (MultiD)

Table 1: ScorecardTM panel of genes Control **Ectoderm** COL₂A₁ CABP7 ABCA4 **ACTB** DRD4 CLDN₁ BMP10 **CTCF** CPLX2 CDH₅ EP300 LMX1A ELAVL3 CDX2 SMAD1 MAP2 **EOMES** COLEC₁₀ ESM₁ MYO₃B FOXA1 Self renewal NOS2 FOXA2 CXCL5 NR2F1/NR2F2 FOXP2 FOXF1 DNMT3B GATA4 NR2F2 OLFM3 GATA6 HAND1 HESX1 IDO₁ **HHEX** HAND2 **PAPLN** LCK PAX3 HMP19 NANOG HOPX PAX6 HNF₁B POU₅F₁ POU₄F₁ HNF4A IL6ST SOX2 **PRKCA** KLF5 NKX2-5 TRIM22 SDC2 LEFTY1 LEFTY2 NR5A2 SOX1 TRPM8 NODAL PHOX2B **PDGFRA** WNT1 POU₃F₃ **PLVAP** ZBTB₁6 PTHLH PRDM₁

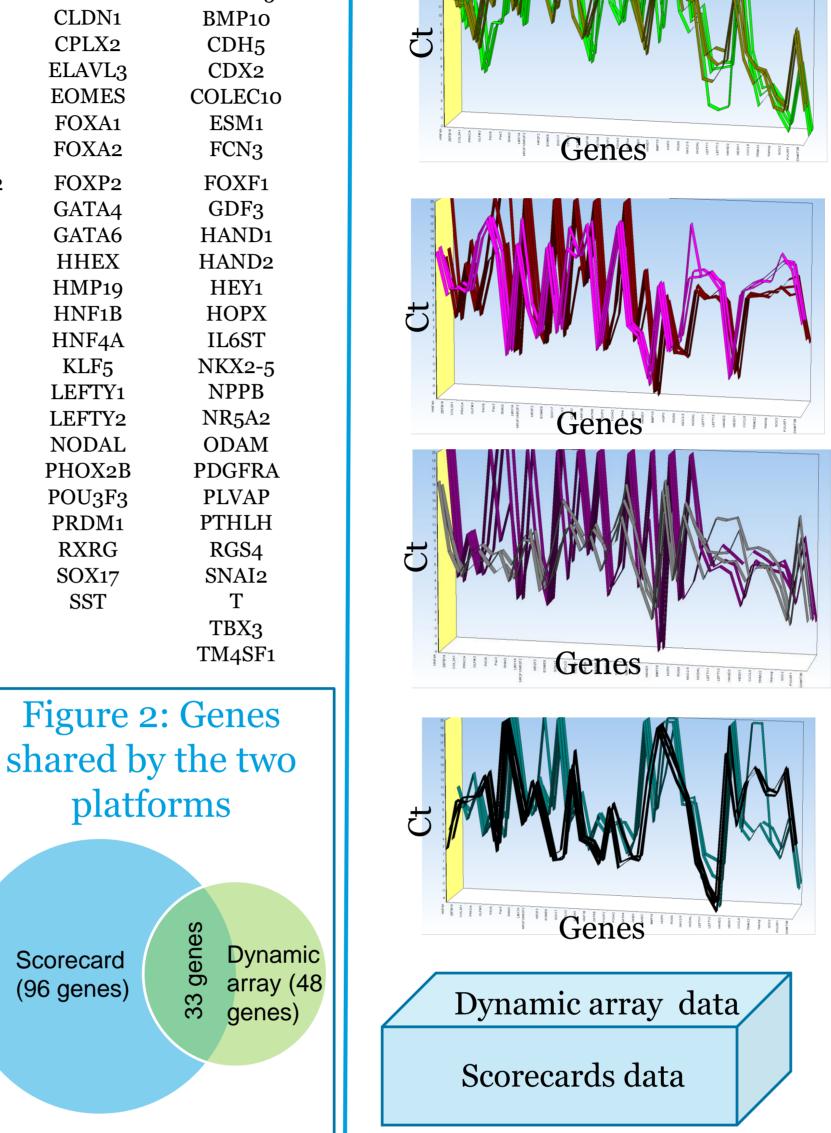
Dynamic array and ScorecardTM RGS4 **RXRG** Figure 1: Differentiation analysis by Scorecard™ and dynamic arrays SOX17 SNAI2 ScorecardsTM TM4SF1

Figure 2: Genes

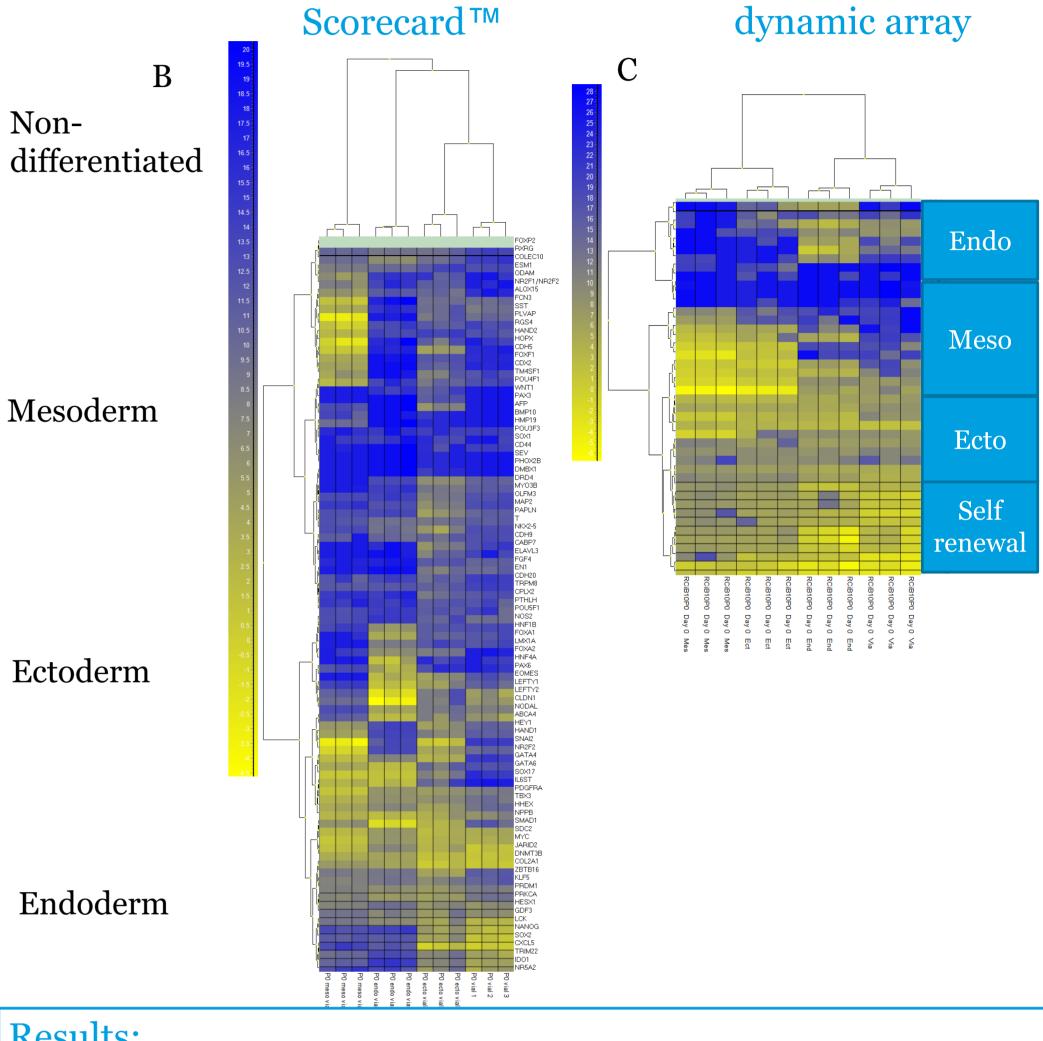
platforms

Scorecard

(96 genes)

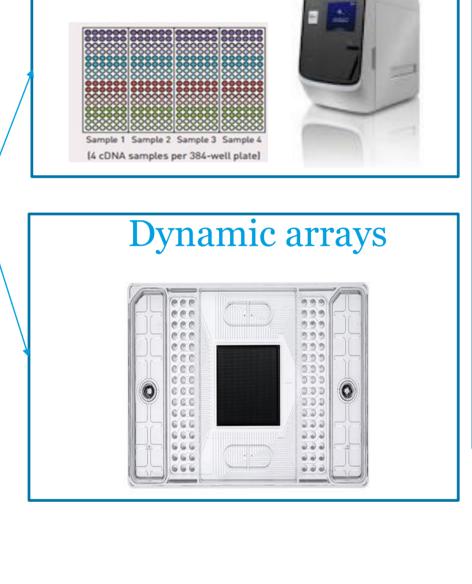


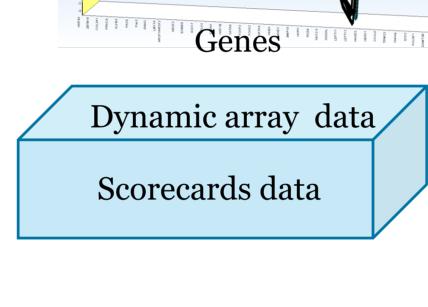
Trend Analysis



Hierarchical clustering

RCiB₁₀ Harvest and iPS extract RNA Non diff Day o cDNA Meso Day 6 Day 6 Ecto Day 6 Endo





Results:

Non-

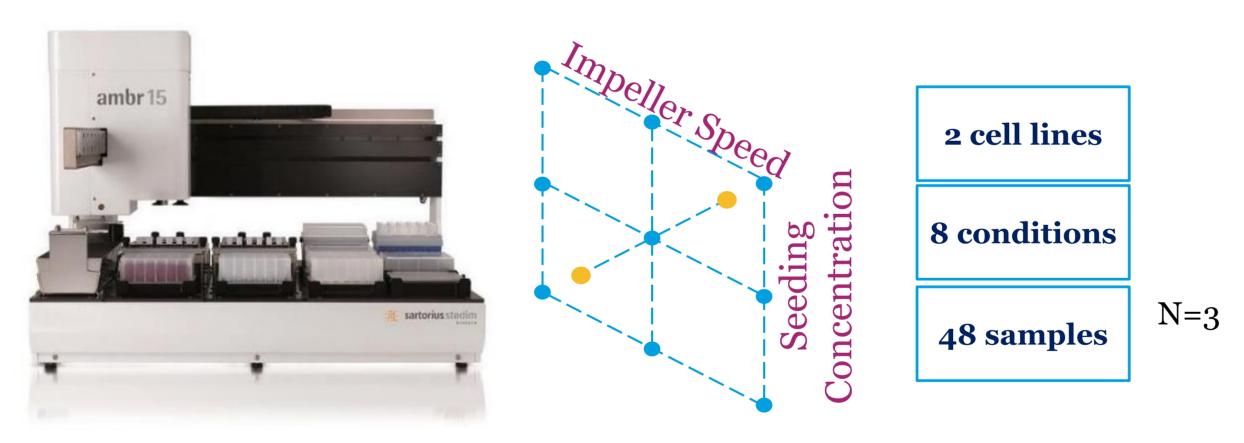
Gene expression trend was compared according to normalised cycle threshold values read by each of the platforms. Both dynamic arrays and ScorecardsTM show high level of similarity, differences between the platform are seen only in non-expressing genes, where there is a different read-out to each of the platforms (Figure 3 A). Hierarchical clustering was used to identify groups of genes that are expressed by differentiated and non-differentiated cells showing using both platforms achieve the same clustering of samples, in accord with their differentiation (Figure 3 B and C).

Comparison of a dynamic array gene expression panel against ScorecardsTM during PSCs culture optimisation

Methods

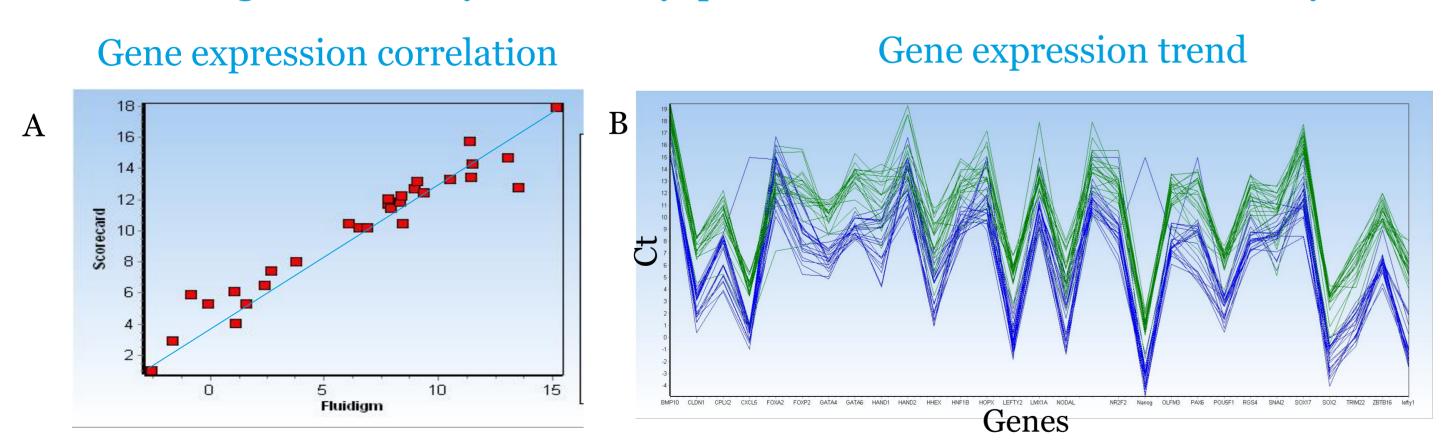
Figure 4: Multi-component bioreactor optimisation

N=3



Design: Culture conditions were optimised on AMBR 15 bioreactor (Figure 4). The effect of culture conditions such as impeller speed, media type and seeding concentration on gene expression was analysed by hPSC Scorecards™ (Life Technologies) and dynamic arrays and the two platforms were compared (Figures 5 and 6).

Figure 5: CGT dynamic array panel vs Scorecard™ correlation analysis

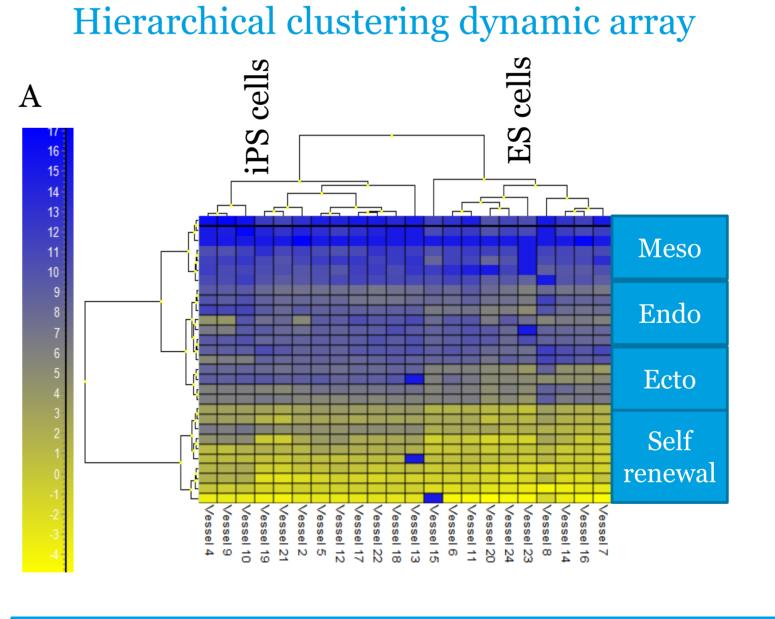


Results: Gene expression trend was compared according to normalised cycle threshold values read by each of the platforms. The trend of gene expression shows correlation (Figure 5 A). Both dynamic arrays and Scorecards™ show high level of similarity in gene expression, however the raw data reads on dynamic arrays show overall lower Ct values compared to Scorecards due to differences in the methods (Figure 5 B) however the trend between different genes remains similar.

Figure 6: CGT dynamic array panel vs ScorecardTM comparison

Figure 3: A comparison of differentiation analysis by ScorecardTM and dynamic arrays

Hierarchical clustering



Results: In-order to compare Scorecards and dynamic arrays, gene expression data was analysed by hierarchical clustering according to normalised Ct values of dynamic array (Figure 6 A) and Scorecards™ (Figure 6 B). Data obtained from both platforms created informative clusters that differentiated between inducible pluripotent (iPS) cell line and Embryonic Stem cell line (ES). In the analysis of both platforms self-renewal genes were highly expressed and there was no differentiation.

Conclusions:

- and Scorecards give Both dynamic arrays comparable results (Figures 3-6);
 - Dynamic array can be applied as in process control for testing multiple culture conditions of PSCs;
- Dynamic array assay is flexible and adaptable; The tested panel will be further extended and we are currently testing additional markers;

