

1 **CMPortal: a community-oriented, data-driven resource to inform protocol design for cardiac modelling from**
2 **human pluripotent stem cells**

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16

17 **Abstract**

18 Protocol design and benchmarking is central to optimising model development using human pluripotent stem cell
19 derived cardiomyocytes (hPSC-CMs). By applying data mining to decades of research and hundreds of peer reviewed
20 studies, we evaluate how protocol variables associate with common properties of cardiac functional and physiological
21 maturation. This resource is publicly accessible through CMPortal, a community-oriented website that provides data-
22 driven tools for researchers to navigate leverage decades of knowledge for benchmarking protocol designs and outcomes
23 for their dedicated applications in developmental biology, disease modelling, and drug screening.

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37 Human pluripotent stem cell-derived cardiomyocytes (hPSC-CMs) are a widely used model for cardiac research,
38 enabling studies of development, physiology, disease mechanisms, and drug responses.¹ However, maturing hPSC-CMs
39 toward adult-like phenotypes remains an ongoing challenge for modeling adult cardiac biology.² Despite development
40 of maturation strategies involving specialized media, co-culture systems, and various types of matrix and biomaterials,
41 optimizing protocol conditions suitable for modelling adult heart phenotypes remain challenging due to high variability
42 in experimental design and outcomes.

43 Ewoldt et al.^{3,4} recently reviewed 300 study designs and demonstrate that while maturity outcomes have improved over
44 time, no significant differences exist between maturation strategies. They highlight difficulties in parsing individual
45 effects of protocol variables due to complex design variables and inconsistent reporting, drawing attention to the need
46 for data-driven approaches and community-oriented benchmarking strategies to improve the accuracy and efficiency of
47 hPSC-CM protocol design.

48 Here, we develop a data-driven platform that offers user-guided analyses of protocol features and their association with
49 functional outcomes. Rather than assuming direct correlations exist between protocol factors and physiological
50 maturation, we use an unsupervised approach that enables detection of bidirectionality and context-dependency allowing
51 parameters to be independently assessed.

52 We first expanded the Ewoldt database³ to 322 studies, including 14 metabolic maturation studies and 8 publications to
53 capture recent developments in the field. We also added protocol design features including cell line sex, ancestry, and
54 transcriptomic markers of myofilament isoform switching. The final database has standardized nomenclature across 400
55 protocol features spanning five categories: experimental design variables (e.g., media composition factors, plating
56 densities, matrix), analysis methods (e.g., imaging techniques, electrophysiology approaches), cell profiles (details about
57 cell line including sex, ancestry, and source), study characteristics (e.g., publication year, journal), and measured
58 outcomes (e.g., measured sarcomere length, contractile force, calcium handling, myofilament isoforms) (**Figure 1A**).

59 We applied a hybrid strategy combining conventional statistics and data mining to assess protocol features against 117
60 target parameters, such as 18 maturity indicators, cell profiles, and applications in disease and pharmacological modeling
61 (**Tables S1 and S2**). Numerical measurements were binned into quantiles (e.g. Q1, Q2, etc.) to enable analysis of
62 protocols ranked by their maturity outcomes. One-hot encoding was used to derive a data format suitable for machine
63 learning, which also accounts for the significant missing data and assumes protocol variables, when not used or reported,
64 are still potentially causal.

65 For data mining, we used random forest classifiers with 10,000 permuted feature sets to associate protocol designs with
66 target parameters through Shannon entropy and Gini impurity, allowing the detection of statistically robust associations
67 even in small and sparsely reported datasets (**Figure 1A** and **Table S3**). Dimensionality reduction via UMAP
68 demonstrated high protocol heterogeneity across studies, consistent with reported findings³ (**Figure 1B**). Despite this
69 heterogeneity, polynomial regression confirmed that the database sufficiently captures protocol features for target
70 endpoints relating to best and worst performing quantiles in a data-limited setting (**Figure 1C**). Quality checks using
71 Jaccard index analysis confirmed that data transformation preserved feature distinctiveness (**Figure S1**).

72 Data mining revealed alignment between protocol variables and expected biological impacts on cardiomyocyte
73 maturation (**Figures 1D-E**). For example, Q1 contractility protocols were significantly associated with Wnt modulation,
74 fibroblast inclusion, and matrix stiffness which are well established factors impacting sarcomere assembly and force
75 generation.^{5,6} Importantly, we identified that identical maturation strategies could be linked to both best and worst
76 maturation outcomes, demonstrating that this approach identifies context-dependent effects of protocol variables and
77 not simply assume their generalized effects (**Figure 1F**). Specifically, the findings suggest that maturation parameters
78 can be independently modulated by distinct sets of protocol variables, challenging the assumption that improving one
79 cardiac maturity metric necessarily enhances all others.

80 We next used a spearman correlation analysis of all continuous variables in the database (n=31) to assess whether
81 maturity measurements recapitulate expected physiological relationships (**Table S4**). Indeed, the data demonstrates that
82 across hundreds of peer-reviewed studies, there are significant positive correlations between expected parameters
83 including resting membrane potential and conduction velocity ($p=1$, $p<0.01$), as well as contractile force and 3D tissue

84 size ($p=0.33$, $p<0.05$) (**Figure 1G**). Interestingly, the database also identifies protocol correlations that provide important
85 considerations in co-culture design. Different cell type compositions revealed positive, null, and negative effects on
86 maturity indicators including calcium kinetics and sarcomere length (**Figure 1H**).

87 We integrated these findings into a community-oriented web resource, *CMPortal* (<https://palpantlab.com/cmportal>). The
88 Database Viewer enables searching of the 322 curated protocols based on five feature categories using techniques from
89 simple keyword matching to multi-criteria searches, with options to toggle column visibility, export customized datasets,
90 and evaluate relevant protocols of interest (**Figures 2A-C**). The Variable Search enables researchers to identify studies
91 and protocols based on topic areas and/or sets of protocol variables which enables unsupervised discovery of relevant
92 protocols based on desired and/or associated protocol features, even when the published protocols do not report the
93 selected metrics (**Figure 2D**). The Protocol Benchmarking tool enables researchers to evaluate differentiation and
94 maturation protocols against database standards. By uploading or manually selecting protocol features and the user's
95 own experimental measurements, the tool compares protocols through a radar chart that highlights maturity indicators
96 across multiple categories and visually distinguishes between predicted, normal-range, and outlier experimental values
97 (**Figures 2E-F**). Lastly, the Enrichment Browser allows researchers to identify protocol features significantly associated
98 with specific outcomes which serves as the data foundation for the search and benchmarking features. (**Figure 2G**).

99 We aimed to validate an unexpected finding from the database showing that contractility and sarcomere length are not
100 correlated maturation endpoints (**Figure 2H**), each having distinct protocol variable enrichments governing their
101 maturation (**Table S2**). This finding suggests that the Frank-Starling effect, one of the most fundamental principles
102 governing the relationship between heart pump function and load, develops through independent mechanisms. We
103 investigated this by first identifying enriched protocol characteristics of Q1 and Q5 contractility, choosing backbone
104 media compositions involving DMEM+fatty acids (FA) (Q1 contractility) and RPMI+B27 (Q5 contractility) (**Figure**
105 **2I**). DMEM+FA confers enhanced contractile maturity due to the high calcium concentration⁷ and fatty acids⁵ that
106 mimic postnatal development. As shown in data mining, impedance-based measurements showed significantly higher
107 peak contraction amplitude in DMEM+FA versus RPMI+B27 (**Figure 2J**), which was also supported by RNA-seq data
108 showing higher metabolic activity from glycolysis to mature fatty acid oxidation in DMEM+FA cells (**Figure 2K**).
109 Consistent with the database correlation, contractility changes occurred without significant variation in sarcomere length
110 (**Figure 2L**). This supports our observation that fundamental features of cardiac maturation, including interdependent
111 mechanisms controlling heart function like the Frank-Starling relationship, are not necessarily mechanistically linked in
112 development. *CMPortal* allows for unsupervised discovery of these context-dependent protocol variables, enabling
113 users to independently optimize protocols for specific functions.

114 *CMPortal* represents the first quantitative approach to overcome significant analytical difficulties of unreported, sparse
115 protocol data, successfully associating context-dependent protocol variables with cell modelling outcomes. These
116 findings challenge a fundamental but rarely challenged assumption that hPSC-CM maturity advances uniformly across
117 all metrics. Instead, the results support a modularised paradigm to develop protocols through selected metrics based on
118 experimental goals, aligning with recent literature demonstrating that attributes of hPSC-CM maturation can be
119 compartmentalized.⁸ We translate these findings into user-friendly tools provided on a public website for researchers to
120 evaluate, design, and benchmark protocols. *CMPortal* provides a streamlined approach to literature review, inform
121 experimental design, and facilitate protocol benchmarking for understanding cardiac biology, diseases, and drug
122 responses as a community-oriented resource for hPSC-CM research.

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131 **Declaration of interests**

132 N.J.P is cofounder and advisor for Infensa Bioscience, a biotechnology company focused on the development of peptide
133 therapeutics for stroke and heart attack. J.E.H. is a cofounder, scientific advisor, and holds equity in Dynamics, a
134 biotechnology company focused on the development of heart failure therapeutics. J.E.H. is a co-inventor on patents for
135 bioengineered human cardiac tissue patches. J.E.H. is co-inventor on licensed patents for engineered heart muscle, and
136 patents relating to cardiac organoid maturation and therapeutics.

137 **Figure 1. *CMPortal* database construction and hybrid data mining approach.**

138 (A) Schematic overview of the workflow for integrating and analysing 322 hPSC-CM studies to evaluate protocol
139 designs and target parameters.

140 (B) UMAP visualizations project the database by five feature categories, enabling assessment of variability in protocol
141 components and identification of shared trends, with annotations shown in color.

142 (C) Relationship between the number of significant enrichments and the number of training protocols. A polynomial
143 model fitted to the best and worst quantiles of maturity indicators (pink and purple) suggests a plateau at 20–30 out of
144 413 protocol features significantly associated with maturity metrics. Grey indicates target parameters for intermediate
145 maturity quantiles, cell profile, and study characteristics.

146 (D–E) Bar plots showing protocol feature metadata for the highest (D) and lowest (E) contraction force quantiles, based
147 on enrichments with $p < 0.025$. Yellow bars indicate the number of studies reporting the relevant variables; purple bars
148 represent the confidence in true feature enrichment. Log-odds ratio (logOR) and Fisher's exact test (FET) estimate
149 overall directionality of enrichment. Co-reporting bias is calculated as the percentage of enriched features co-published
150 more often than expected by chance, using the mean Jaccard Index (JI) distribution for a matched number of features.

151 (F) Stacked bar plot compares maturation strategies revealing significant context-dependent effects on maturation
152 properties.

153 (G) Table summarizing representative Spearman correlations among numerical protocol features, showing expected
154 associations (full data in Table S4).

155 (H) Spearman correlation plots of cardiac cell ratios in 3D cultures show opposing relationships with sarcomere length
156 and calcium relaxation time as maturity indicators.

157 **Figure 2. *CMPortal* website interface and experimental use cases.**

158 (A) Database Viewer allows users to explore 322 hPSC-CM protocols across 483 protocol features, and to install the
159 filtered database.

160 (B) Pie charts of the database show media and substrate usage of metabolic maturation protocols.

161 (C) Example UMAPs show segregation of studies by the protocol feature categories. Cell line ancestry shows distinct
162 trends of protocol variable and analysis method usage. Common media, like most features, show no clear trends.

163 (D) Variable Search enables protocols retrieval by any desired and/or significantly enriched features by target parameters.

164 (E) Protocol Benchmarking module enables users to upload and compare protocols against the database using
165 experimental data and reference ranges estimated with the feature enrichments.

166 (F) Radar plot generated by *CMPortal* for comparing maturation outcomes between Protocol 80 and 250 which are
167 amongst the most cited studies in the database. The plot shows protocol 80 being associated to properties of higher
168 transcriptional and contractile maturation, but lower electrophysiological and calcium handling maturation. Hollow
169 datapoints are estimated with enrichments when no experimental data is available.

170 (G) Enrichment Browser retrieves enrichments across the 117 target parameters.

171 (H) Spearman correlation of pairwise, co-reported values of contractility and sarcomere length in 322 protocols. No
172 significance was found.

173 (I) *CMPortal*-informed experiment designed to compare the impact of media composition on contractility.

174 (J) Representative traces and bar plot of impedance-based contractile measurements in hPSC-CM across time show
175 higher peak amplitude for DMEM+FA and RPMI+B27 cells.

176 (K) Bulk RNA-seq analysis of pathway-specific gene panels indicates metabolic reprogramming towards higher
177 maturity in hPSC-CMs cultured in DMEM+FA compared to RPMI+B27.

178 (L) Representative alpha-actinin staining (green), DAPI (blue) in fixed hPSC-CM confirms independence of
179 contractility from sarcomere length. Bar charts values are mean \pm standard error whiskers with student's *t*-test
180 significance (G, I). *P<0.05, **P<0.01, ***P<0.001, ****P<0.0001, ns: non-significant. Study Char. : Study
181 Characteristic; Meas. Endp.: Measured Endpoint; Sarco. Length: Sarcomere Length.

182 **References**

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A

Database Processing

300+22 published hiPSC-CM studies

Terminology Unification

Additional Protocol Features

Additional Variable Categories

One-hot Encoding

Database reveals:

>483 protocol features

322 protocols

1 Common trends

2 Co-reported features common in studies

3 Traditional Statistical Tests

Initial evaluation of feature relationships

4 Data Mining Approach

Permutation-based importance testing

4 Data Mining Approach

Permutation-based importance testing

10k Permuted Datasets

CMPortal Database

Null distributions of feature importances

Significant features scored against null distribution

CMPortal Database

Relates protocol design to maturation, physiological, pharmacological, and more applied endpoint measurements

Relationship between protocol design and cell application/outcome

4 3 Traditional Statistical Tests

Initial evaluation of feature relationships

4 2 Co-reporting bias

Jaccard index assesses enrichments arising from co-reporting in studies

Features

Enrichments

Protocols

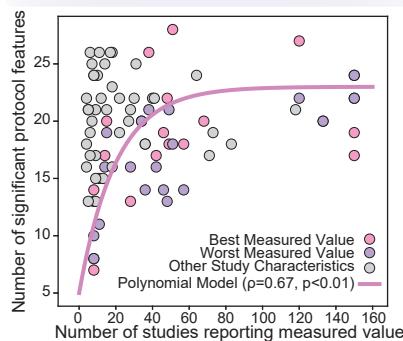
3 Odds Ratios (logOR)

Fisher's Test (FET)

Calculate significance and direction of association

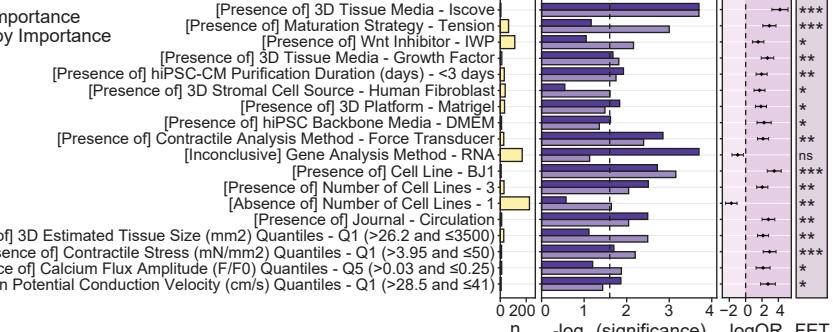
but suffers from missing data, high variability, low power

C 4 Overview of the number of revealed associations and training sample size



D 4 3 Protocol Features Associated with Contractility (mN) Q1 (>1.04 and ≤20) [10]

Gini Importance
Entropy Importance

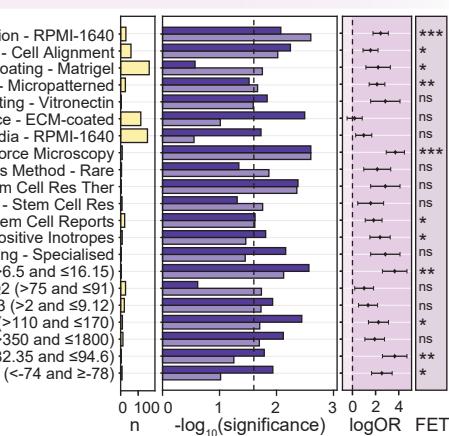


Co-reporting Bias: 15% (p<0.001)

E 4 3 Protocol Features Associated with Contractility (mN) Q5 (>0 and ≤0.04) [10]

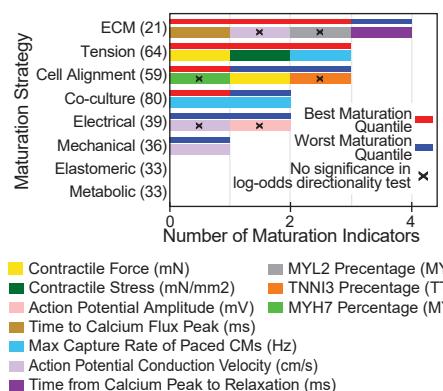
Gini Importance
Entropy Importance

[Presence of] New Media for Maturation - RPMI-1640
[Presence of] Maturation Strategy - Cell Alignment
[Presence of] hiPSC Matrix Coating - Matrigel
[Presence of] 2D Surface - Micropatterned
[Inconclusive] Coating for Replacing - Vitronectin
[Inconclusive] 2D Surface - ECM-coated
[Inconclusive] hiPSC-CM Maturation Media - RPMI-1640
[Presence of] Contractile Analysis Method - Traction Force Microscopy
[Inconclusive] Gene Analysis Method - Rare
[Inconclusive] Journal - Stem Cell Res Ther
[Inconclusive] Journal - Stem Cell Res
[Presence of] Journal - Stem Cell Reports
[Presence of] Pharmacological Intervention - Positive Inotropes
[Inconclusive] Disease Modelling - Specialised
[Presence of] TNNI3 Precentage (TNNI1) Quantiles - Q2 (>0.5 and ≤16.15)
[Inconclusive] 3D CM Ratio (CM-EC-SC) Quantiles - Q2 (>75 and ≤91)
[Inconclusive] 3D Estimated Tissue Size (mm2) Quantiles - Q3 (>2 and ≤9.12)
[Presence of] Action Potential Amplitude (mV) Quantiles - Q1 (>110 and ≤170)
[Inconclusive] Cell Area (um2) Quantiles - Q3 (>350 and ≤1800)
[Presence of] MYH7 Percentage (MYH6) Quantiles - Q1 (>82.35 and ≤94.6)
[Presence of] Resting Membrane Potential (mV) Quantiles - Q2 (<-74 and ≥-78)



Co-reporting Bias: 10% (p=0.02)

F 4 3 Maturation Strategies Compared

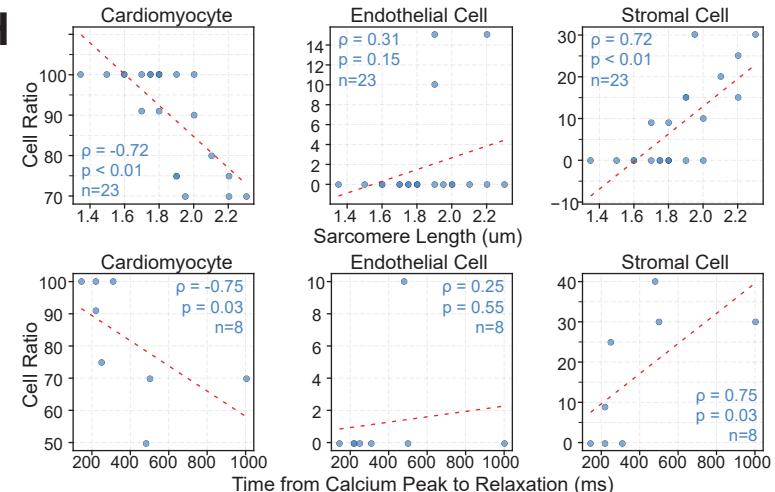


Best Maturation Quantile
Worst Maturation Quantile
No significance in log-odds directionality test

G

Correlated Protocol Features	Spearman p	pval	n
2/3D	Resting Membrane Potential (mV)	1.00	>1.00E-04
	Conduction Velocity (cm/s)		4
	Seeding Confluence (%)	0.63	8.35E-03
	Contractile Stress (mN/mm2)		16
	Wnt Inhibitor Duration (days)	0.76	3.00E-02
3D	MYH7 Percentage (MYH6)		8
	Beat Rate (bpm)	-0.78	2.14E-02
	Insulin Withdrawal Duration (days)		8
	Conduction Velocity (cm/s)	-0.76	4.57E-02
	Insulin Start Day		7
3D	Contractile Force (mN)	-0.23	0.41
	Sacromere Length (um)		15
	Differentiation Purity (%)	-0.37	1.39E-03
	3D Estimated Cell Density (mil cells/mL)		70
	3D Estimated Tissue Size (mm2)	0.33	4.02E-02
3D	Contractile Force (mN)	0.33	4.02E-02
	3D CM Ratio (CM-EC-SC)	-0.71	9.94E-04
3D	Resting Membrane Potential (mV)		18
	3D CM Ratio (CM-EC-SC)		

H



A 1 Database Viewer

Search protocols most relevant to a target parameter

Explore 483 protocol features across five categories

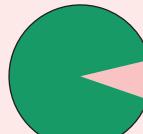
	Protocol Variable	Analysis Method	Cell Profile	Study Characteristic	Measured Endpoint
DOI 1	Maturation Strategy	Gene Analysis	hiPSC Cell Line	Disease Modelling	Sarco. Length
DOI 2	Metabolic Maturation	PCR	BJ1	Ischaemia	N/A
DOI 3	Metabolic Maturation	Gene Analysis	iCell	Ischaemia	Q2
DOI 4	Metabolic Maturation	PCR	BJ1, iCell	Ischaemia	Q1
...
DOI N	Metabolic Maturation	PCR	BJ1, iCell	Ischaemia	Q5

Review all 322 studies

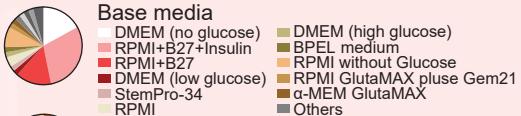
Sort or filter protocols by features of interest

Numerical measures are binned into quantiles

B Understand media composition of selected studies

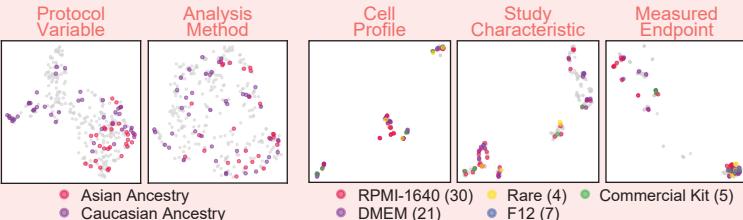


322 hiPSC-CM Studies
Metabolic Maturation



Metabolic substrates
Fatty Acid and Lipids
Sugar and Carbohydrates

C Assess trends by protocol components with UMAP visualisation



D 2 Variable Search

Toggle features of interest with ease on CMPortal

Search target: Ischaemia Modelling
Filter results by: IWP2

Search by Matching... Target's Enrichments Protocol Features

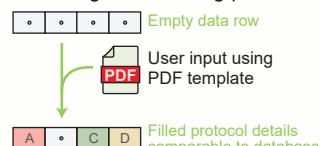
Feature categories containing enrichments for the search target

Score	Target	Filter(s)	Other Features	Enrichment found in protocol:			
				Protocol Variable	Analysis Method	Cell Profile	Study Char. Meas. Endp.
DOI 1	95	True	IWP2	✓	✓ ✓
...	92	True	...	✓	...	✓	...
...	89	Not Reported	...	✓	...	✓	✓
...
DOI n	0	False	...	✓	...	✓ ✓	...

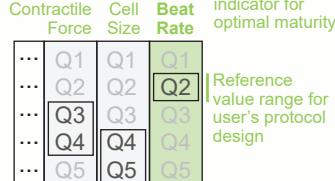
Search engine can be toggled, searching by enrichments to identify relevant protocols designs regardless of their intended application and reported values

E 3 Protocol Benchmarking

Compare custom protocol designs to existing protocols



Maturity indicator comparison:



G 4 Enrichment Browser

Query CMPortal to inform testable hypotheses for experimental considerations and designs

Select target...
Maturation Indicators
Physiological Descri..
Cell Sex and Ancestry
Disease Modelling
Pharmacological Test

Contractile Force
Ca Flux Amplitud..
MYH7 Percentage
Action Potential ...

Query Outcomes:
Associated with BEST (Q1) maturation:
hiPSC Backbone Media - DMEM
Cell Line - BJ1
...
Associated with WORST (Q5) maturation:
Maturation Media - RPMI-1640
Maturation Strat. - Cell Alignment

Day	Contractile Impedance (Ω)								Gene Expression							
	-1	0	1	2	3	4	5	6	7	8	RPMI+B27	DMEM+FA	RPMI+B27 and DMEM+FA	Maturation Media Comparison	RPMI-1640	DMEM no glucose/10 mM HEPES
Media	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI+B27	RPMI-1640	DMEM no glucose/10 mM HEPES
Base Media	RPMI-1640	2% B27 with insulin	2mM L-carnitine	5mM creatine	5mM taurine	1mM nonessential amino acids	1x insulin-transferrin-selenium	1x linoleic-oleic acid								
Supplements																

