A Mechanical Model of Early Somite Segmentation

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Abstract:

The clock-and-wavefront model (CW) hypothesizes that the formation of body segments (somites) in vertebrate embryos results from the interplay of molecular oscillations with a wave traveling along the body axis. Our experiments, however, suggest that somites self-organize mechanically during a wave of epithelialization of the dorsal pre-somitic mesoderm that precedes somite segmentation in chicken embryos. Signs of somite formation first appear in the dorsal pre-somitic mesoderm, as a series of arched segments of epithelial monolayer separated by clefts. The clefts seem to correspond to the position of the eventual inter-somite boundaries. We develop a corresponding mechanical model of epithelial segmentation in early somitogenesis in which a wave of apical constriction leads to increasing tension and periodic failure of adhesion junctions within the dorsal epithelial layer of the PSM, forming clefts which determine the eventual somite boundaries. Computer simulations of this model can produce spatially periodic segments whose size depends on the speed of the contraction wave (W) and the rate of increase of apical contractility (Λ). We found that two values of the ratio Λ/W determines whether this mechanism produces spatially and temporally regular or irregular segments, and

whether segment sizes increases with the wave speed (scaling) as in the clock-and-

wavefront model. Based on this, we discuss the limitations of a purely mechanical model

of somite segmentation and the role that biomechanics play along with CW during

somitogenesis.

Keywords: Somitogenesis, biomechanics, mesenchymal-to-epithelial transition

INTRODUCTION

Somitogenesis in vertebrate development sequentially and periodically creates

metameric epithelial balls (somites) along the elongating embryo body from two rows of

undifferentiated, loosely connected mesenchymal rods of cells called the pre-somitic

mesoderm (PSM) while new PSM cells are continuously being deposited from the tail

bud at the caudal (tail) end of the embryo (*Pourquié, (2001)*). At any given rostral-

caudal position, a pair of nearly equal-sized somites form simultaneously on both sides

of the neural tube, between the ectoderm and the endoderm. These transient structures

are the precursors of vertebrae, ribs and many skeletal muscles, with many birth defects

associated with a failure in one or more steps in this long developmental process (*Christ*

et al, (1995)).

Somitogenesis is strikingly robust to perturbations (both spatial and temporal).

Changes in the total number of embryonic cells or in the rate of new cell addition at the

caudal end lead to compensating changes in the size and timing of somite formation so

that the embryo eventually produces the same final number of somites as in normal

development (*Cooke, (1995)*). One way to achieve this conservation of final somite

number is for the size of the somites to increase linearly (*scale*) with the speed of the

caudal-moving position of the determination front or with the rate at which cells join the

caudal end of the PSM.

Models seeking to explain somite formation include the 'cell cycle model,' which couples

cells' somite fate to their cell-cycle phase (Stern et al (1988), Primmett et al (1988)) and

reaction-diffusion models (*Meinhardt (1982), Cotterell et al (2015)*). Currently, the most

widely accepted family of models employ a 'clock and wavefront' (CW) mechanism

which combines caudally progressing waves of determination and differentiation with an

intracellular oscillator which determines cell fate based on its phase at the moment of

determination (*Cooke et al (1976)*). Following the identification of the first oscillating

protein in the PSM (*Palmeirim et al (1997)*), many computer simulations, of varying

complexity have implemented different CW models. Most CW models reproduce the

experimentally observed scaling of somite size with clock period, wavefront speed and

rate of elongation of the PSM (*Hester et al (2011), Tiedemann et al (2012), Baker et al*

(2006))

Recent experiments have shown that somite-like structures can form without either a

clock or a wavefront (*Dias et al (2014), Cotterell et al (2015)*)). The ability of somites to

form without clock or wavefront suggests that we should consider other mechanisms

that could lead to spatially and temporally periodic sequential division of the PSM into

regular segments. Recent experiments show that applied tension along the rostral-

caudal axis can induce the formation of intersomitic boundaries in locations not

specified by clock-and-wavefront signaling (*Nelemans et al (2017)*), suggesting that

mechanical mechanisms may be important in generating inter-somitic boundaries.

In 2009, Martins and colleagues imaged the morphology of cells during chicken

somitogenesis *in vivo*, showing that cells elongate, crawl and align with each other as

they form a somite (*Martins et al (2009)*). Cells in the PSM begin to change their

behaviors gradually from mesenchymal to epithelial long before the formation of visible

somites. Cells epithelialize gradually during somite formation, with epithelialization

beginning long before segments separate from each other. Our recent experiments

show that in addition to the classical rostral-caudal progression of epithelialization,

epithelialization begins along the dorsal surface of the chicken PSM and propagates

ventrally. As cells approach the time of the physical reorganization which generates

somites they gradually increase the density of cell-cell adhesion molecules at the cell

surface (*Duband et al (1987)*) and decrease their motility (*Bénazéraf et al (2010)*).

When somites form, these gradual changes lead to a discontinuous change, somewhat

like a jamming transition, where the tissue locally switches from extended and liquid-like

to condensed and solid-like (in zebrafish) and the cells in each forming separate from

those in the neighboring somite (*Mongera et al (2018)*).

Based on earlier observations of periodic (metameric) organization in the rostral-caudal

direction before somite formation (which *Meier (1979)* and *Tam et al (1982)* called

somitomeres) we performed Scanning Electron Microscopy (SEM) of chicken embryos

fixed at various stages of somitogenesis. These 3D SEMs consistently show that the

dorsal monolayer of PSM cells begins to epithelialize and form a series of arched tissue

three to four somite lengths caudal to the forming somite (Figure 1A). These cohorts are

about the size of somites suggesting that they define the groups of cells that will form

particular segments a few hours later.

These observations lead us to ask whether spatially and temporally periodic tissue

segmentation and somite boundary positioning could result from a mechanical instability

mechanism in the dorsal pre-somitic epithelium. We investigate a model in which a

continuous caudally-travelling wave of apical constriction in the pre-somitic dorsal

epithelium can lead to the sequential formation of periodic, spatially separated tissue

segments. Our 2D simulation of a cross-section of the epithelial monolayer under this

model can produce either constant-size segments or segments whose size increases

(*scales*) with wave speed and inverse rate of increase of apical contractility. A roughly

constant critical ratio of the wave speed to apical contractility build-up rate defines the

boundary between these two domains. A second critical ratio also predicts when this

mechanism produces spatially and temporally regular segments from irregular

segments.

Early signs of boundary specification

To observe cell morphologies during somite formation, we fixed and sectioned sagittally

chick embryos using a guillotine knife to produce planes of section almost perfectly

aligned to the embryonic midline along the entire body axis. We examined the embryos

using Scanning Electron Microscopy (SEM). Preliminary observations (*Stern et al*

(2015)) reveal that the dorsal region of the PSM starts to organize into a monolayer

comprised of progressively elongated (columnar) cells as early as 4-5 somite-lengths

caudal to the last-formed somite (S1).

Higher magnification SEM images of the PSM for a few somite lengths caudal to the

forming somite reveal roughly periodic (in the rostral-caudal direction) clusters of cells of

about 12 cells in length along the caudal-rostral axis(Figure 1A). Cells within these

clusters are usually wedged shaped, with their apical (ventral facing) sides more

constrited than their basal (dorsal facing) sides, which together gives the whole

segment a dorsally arched aspect. The arched clusters are separated by what we refer

to as clefts, or connecting regions, with cells that are either squeezed, apically

separated, basaly constricted or non-elongated (arrowheads in Figure 1A). The

distances between sequential clefts and between the first cleft and the S1 somite is

regular and roughly corresponds to the caudal-rostral length of the future somites.

The observed position of the clefts between arched cell cluster closely corresponds to

the posterior boundaries of the next somites to form (S-1,S-2 and S-3), suggesting that

these structures arise from a pre-patterns mechanism of the future inter-somite

boundary. At the same time, the overall convergence of the cells' apical sides between

these clefts raise the possibility that apical constriction takes place in the pre-somitic

dorsal epithelium as the PSM matures. Based on this we hypothesize a mechanical-

instability model for the segmentation of the dorsal pre-somitic epithelium. The model

has two components: a caudally-travelling wave of apical contractility activation that

leads to a gradual apical constriction of those cells (Figure 1C,D); and a maximum

tension load between neighboring apical sides that determines the appearance of the

clefts and segments the tissue (Figure 1E,F). The model assumes that the cells have an

well-defined apical-basal polarity, and that cells are positionally constrained within a

dorsally aligned monolayer, so that increased apical tension cannot be resolved by

tissue curvature.

Figure 1 – Early signs of pre-somitic epithelium segmentation and conceptual model

(**A**) SEM images of para-sagitally sectioned chicken embryos shows the epithelialization of dorsal PSM

cells about 5 somite-lengths caudal to the S1 somite. Cells are colored according to their aspect ratio, as

shown in (**B**). White arrowheads show the positions of segmentation clefts. (**C-F**) Conceptual model of

mechanical-instability mechanism of tissue segmentation. (**C**) Epithelialized cells, with defined apical (red)

and basal (blue) sides form a rostral (right)-caudal (left) monolayer along the dorsal side of the PSM. (**D**)

A caudal-moving wave of myosin activation (orange arrow) initiates apical constriction of the cells in the

monolayer. (**E**) As more cells begin to constrict apically, regions of high tension in the rostro-caudal

direction develop within the tissue. (**F**) At some point the increasing tension causes the failure of the

adhesion junctions between cells, leading to the appearance of a cleft/tear in the tissue (red double

arrowed line) and the division of the initially continuous PSM into segmented pre-somitic epithelium.

A mechanical model of dorsal PSM pre-patterning for segmentation

Our model is focused on the cell behaviors of the pre-somitic dorsal epithelium. Based

on the SEM observations, we start the simulations when the dorsal PSM cells have

already started to form a monolayer, with defined apical and basal polarity, and an

aspect ratio of 2:1 (Figure 2C). The basal sides of the cells face the ectoderm dorsally

and their apical side faces the more ventral PSM cells. For simplicity we model the

monolayer as single row of cells extended along the AP axis. The cells are aligned and

flanked by each other in this monolayer. We propose that the cells are coupled together

by labile adhesion along their entire surfaces and strong junctional adhesions located

sub-apically (*Shirinifard et al (2012)*). The latter are or special importance as they also

couple the apically localized actin-myosin cytoskeleton of neighboring cells.

As stated before, we hypothesize that the arched segments and clefts observed in this

layer are due to the gradual activation of apical constriction along those cells. Myosin

activity at the cells' apical side can generate contractile forces that reduce their apical

surface area, give cells a wedge shape and bring together neighboring cells (Figure 1D)

(*Baker et al (1967), Pilot et al (2005), Martin et al (2014), Sawyer et al (2010), Pearl et*

al (2017)). Furthermore, if apical contractile forces exert too much load on junctional

adhesion sites, tissue tears appear, as observed in *Drosophila* embryos (*Martin et al*

(2010)). Similar to a proposed 'determination front' at around the middle of the PSM (

Cooke et al (1976), Dubrulle et al (2001)), we postulate a complementary, posteriorly

travelling activation wave that induces apical constriction along the maturing PSM. As

the wave passes, cells slowly increase their apical tension, eventually leading to the

breaking of contacts between neighboring cells and the formation of a series of clefts

that correspond to the future somite boundaries (Figure 1C-D). The speed at which this

wave passes through the tissue (W) and the rate at which each pair of activated cell

increases their apical contractility (Λ) are key parameters of our model that together

determine the segmentation periodicity.

COMPUTATIONAL MODEL

Our model is implemented using the Cellular Potts (CPM), or Glazier-Graner-Hogeweg

(GGH) model (*Graner et al (1992)*) and simulated with the open-source CompuCell3D

software (*Swat et al (2012)*). In the CPM/GGH framework each cell, or cell region is

spatially represented by a collection of pixels that forms a domain within a fixed grid

(here, rectangular). Cell/domain properties such as size, mobility, adhesion preferences

and distance constraints with other domains, are defined by a Hamiltonian, or energy

cost-function:

(Eq. 1) $$H = H_0 + H_V + H_I + H_A + H_B,$$

where each term will be defined shortly.

The initial configuration of all cell/domains evolve in time by a series of random

neighboring pixel copy attempts whose acceptance is governed by a Metropolis

algorithm, with a number of pixel-flips equal to the grid size defining the time unit of the

simulation, a Monte Carlo Step (MCS).

Similar to what has been done in previously models (*Belmonte et al (2016), Dias et al*

(2014)), each cell is composed of 3 domains representing the apical, basal and core

regions of an epithelial cell. The size of the domains is maintained by a volume

constraint in the Hamiltonian:

(Eq. 2) $$H_{Volume} = H_V = \sum_\sigma \quad \lambda_V (V(\sigma) - V_T(\sigma))^2,$$

where the sum is over all domains (σ), $V(\sigma)$ is the current cell volume, $V_T(\sigma)$ is the cell

target volume, and λ_V is a lagrangian multiplier setting the strength of the constraint.

The cell aspect ratio AR, defined as the ratio of apical/basal length to cell width, is

maintained through the existence of spring-like distance constraints between the center

of mass of the three domains belonging to the same cell:

(Eq. 3) $H_{Internal\ Links} = H_I = \sum_{\sigma,\sigma'} \lambda_I (L(\sigma,\sigma') - L_{IT}(\sigma,\sigma'))^2,$

where the sum is taken over all pairs of domains within the same cell, $L(\sigma,\sigma')$ is the

current distance between the center of mass of the two domains, $L_{IT}(\sigma,\sigma')$ is the

corresponding target distance, and λ_I is the strength of the constraint. To prevent cells

from bending we set the target distance between apical and basal domains equal to the

sum of the target distance between the core domain to the apical and basal domains.

Labile adhesion between cells is modeled with the standard Potts model Hamiltonian

term:

(Eq. 4) $H_{Labile\ adhesion} = H_0 = \sum_{i,j} J(\sigma_i, \sigma_j),$

where the sum is taken over all fourth-order neighboring pairs of grid coordinates i and

j, σ_i and σ_j are the cell domains at grid coordinates i and j, respectively, and J is

the contact energy per unit area between those domains.

Apical constriction is a cell autonomous process that may lead to tissue-level events,

such as invagination, through the coupling of the internal contractile activity of actin-

myosin cytoskeleton of neighboring cells with their adhesion junctions. Since we are

interested only on the tissue level effects of this process, we model junctional adhesion

and apical constriction similar to (Eq. 3), with a distance constraint between neighboring

apical domains:

(Eq. 5) $H_{Apical\ links} = H_A = \sum_{\sigma,\sigma'} \lambda_A (L(\sigma,\sigma') - L_{AT})^2,,$

where the sum is taken over all pairs of neighboring apical domains, $L(\sigma,\sigma')$ is the

current distance between their center of mass, L_{AT} is the target distance, and

$\lambda_A(\sigma,\sigma')$ is the time-varying strength of the constraint. The target distance between

neighboring apical domains is constant throughout the simulations and set to a value

much shorter than the width of the cells (3 pixels). Initially the constraint $\lambda_A(\sigma,\sigma')$,

which we interpret here as the combined strength of apical cytoskeletal activity between

cell pairs, is set very low ($\lambda_A = 20$), so that there is no effect on the tissue or individual

cell shapes. A similar form ($H_{Basal\ links}$) is used for neighbouring basal domains with a

fixed $\lambda_B = 100$. This is to help ensure the top part of the tissue stays together once the

apical compartments have separated.

Above the basal side of cells, there is a collection of domains representing the ectoderm

and fibronectin- and laminin-rich extracellular matrix that progressively forms a basal

lamina (*Rifes et al (2007)*). These domains have a volume constraint and liable

adhesion with the basal side of the epithelial cells that helps to maintain the alignment

of the cells below them, while allowing some upward/downward movement of the cells.

Below their apical sides, cells are adjacent to a loose mesenchyme which we choose to

model as a single domain with no volume constraint and labile adhesion to the cells.

The epithelium is flanked anteriorly and posteriorly by relatively immotile cells, which are

not allowed to change their shape or properties over time. Between them there are

about 115 to 205 cells, depending on the conditions being tested in the simulations.

Apart from the immotile cells that make the AP boundaries, all cells have identical

properties and are allowed to change shape and properties during the course of the

simulation.

Initially all cells are in an inactive state, without exerting significant forces on each other.

We define a 'wave of constriction activation' which moves through the tissue from the

anterior end at a fixed speed W (in units of cell/MCS), so that the next posterior cell is

activated $1/W$ MCS after the lastly activated cell. Apical constriction is implemented by a

linear increase in the strength of apical cytoskeletal activity parameter λ_A between a pair

of activated cells. This variable varies from a base value of 20 up to 600 at a specific

build-up rate $\Lambda = d\lambda_A/dt$ (in units of 1/MCS). As a pair of cells increase their conjugated

apical cytoskeletal activity parameter λ_A, the actual tension between their apical

domains evolves in time, and is defined as:

(Eq. 6) $T(\sigma, \sigma') = -2\,\lambda_A(L(\sigma, \sigma') - L_{AT})$

When this tension value exceeds a maximum threshold $T_{break} = -7500$ the link

between the apical domains of neighboring cells is broken, resulting in the appearance

of a somitic cleft.

Variation in cell aspect ratios was implemented by adjusting the internal distance

constraints between domains in Eq. 3. We also choose to keep the total sum of all

domain target volumes constant and adjust the initial width and length of the cells to

satisfy the internal distance constraints.

Standard Simulation Parameters

Our simulations have 4 key parameters, which we will vary systematically in this study:

1) *the wave speed W*, which travels from the anterior to the posterior end of the

302   tissue and determines when cell pairs will start apical constriction;

303   2) *the rate of increase of apical contractility* (*build-up rate*) $\Lambda = d\lambda/dt$, which

304   determines how fast neighboring cells strength of pull each other apical domains;

305   3) *the apical tension breaking threshold T_{break}*, which determines when the resulting

306   tension (Eq. 6) will lead to the break of neighboring apical links (Eq. 5) and the

307   split of the cell pairs; and

308   4) *the cell aspect ratio AR*, which defines how elongated the cells are at the

309   beginning of the simulation.

311   The standard values of all 4 parameters can be found in (Table 1). A full list of all

312   simulation parameters can be found in Supplementary Material.

Parameter	Symbol	Base value (units)
Wave speed	W	0.003 (cells/MCS)

Apical contractility build-up rate	Λ	0.05 (MCS⁻¹)
Apical tension breaking threshold	T_{break}	-7500 (dimensionless)
Cell aspect ratio (height/width)	AR	2 (dimensionless)

Table 1 - Base values of the 4 key parameters of our model.

Metrics

To analyze the behavior of our model with respect to variation in key parameters we

define and measure the following metrics:

1) *Average Segment Size $< S >$*: calculated by measuring the number of cells

between two consecutive clefts in our simulations. We measure the sizes both

during the course of the simulation - by taking into account the first occurence of

a cleft located posteriorly to the posterior-most cleft - and at the end of the

simulation in order to assess the presence of any splitting events.

2) *Average Distance Delay $< D >$*: defined as the distance between the wave

position at the time of the last occurring cleft. This metric is measured during the

time course of the simulation. This metric is calculated in units of the average

segment size $< S >$ for the corresponding set of parameters.

3) *Average Segmentation Time $< \tau >$*: defined as the time (in MCS units) elapsed

between the appearance of two consecutive clefts.

To avoid boundary effects, all parameters are measured after the appearance of the

first segmentation cleft and before a distance of $3 < S >$ from the last formed cleft to the

posterior boundary.

RESULTS

Tension profiles as a function of number of cells and constriction strength

We first test our model by creating small epithelial monolayers composed of 4 to 16

cells of fixed aspect ratio $AR = 2$ (Figure 2A). The strength of apical contractility of all

cell pairs was increased simultaneously at a fixed build up rate of $\Lambda = 0.05$, from $\lambda_A = 0$

up to a maximum value of $\lambda_A = 600$, without enforcing breakage of apical links. We then

observe the shape of the tissue and the average cell pair tension (Eq. 6) over 20,000

MCS after λ_A has reached the maximum value. As the chain increases in size, cell pairs

in the middle of the segment are under higher tension compared to cell pairs near the

periphery (Figure 2B,C). This forms the premise of our model - as the tissue becomes

larger, more tension is accumulated by the cell pairs which, in the presence of a

maximum tension load, predisposes the tissue to regular segmentation.

Figure 2 - Tissue tension distribution as a function of the strength of apical contractility

(**A**) Snapshots of 4 simulations of 10 cells with different levels of apical contractility strength. The higher

the levels of λ_A the more constricted the tissue becomes. (White - ectodermal tissue; black - non-modeled

PSM; cells domain colors as in Figure 1C-F; vertical white lines - internal distance constraints between

cell domains (Eq. 3); horizontal white lines - distant constraints between domains of neighboring cells (Eq.

5)). (**B**) Plot of average apical tension (Eq. 6) between cell pairs at the end of multiple simulations with

different number of cells (from 4 to 24). In all these simulations λ_A is set to 600. (**C**) Plot of maximum cell

pair tension versus the number of cells in the tissue for different values of λ_A.

Gradual and sequential activation of apical contractility leads to tissue

segmentation

Next we investigated whether sequential activation of apical constriction can slice a

tissue to generate regularly sized segments. From now on, all simulations have a large

number of cells and a posteriorly travelling wave of activation sequentially initiates a

gradual and linear increase of the strength of apical contractility of cell pairs. The wave

speed W and build-up rate of apical contractility Λ will be varied systematically around

their base values (see Table1). In addition, apical cell pairs will have a maximum

tension load T_{break}, after which their links break and a tear may appear between the

apical region of those cells.

Figure 3A shows a time series of a zoomed-in section of a simulation with 115 cells

where 4 tissue segments form. With the exception of the first and last two segments

(due to boundary conditions), all segments are of similar size ($< S >= 11.36 +/- 1.45$)

and clefts form at similar time intervals ($< \tau >= 3442.37 +/- 587.87$) for our standard

set of parameter values (see Table1). We also looked at the evolution of the tension

profile for 3 sequential segments. As the wave passes, the tension between cell pairs

gradually increases, with the anterior-most pair with the higher tension (Figure 3B and

green lines in Figure 3C). After the formation of the anterior cleft/boundary the pattern is

inverted, with the anterior-most cell pairs now more relaxed and the posterior most pairs

now under higher tension (compare green and blue lines in Figure 3C). Formation of the

posterior-most cleft/boundary relaxes the tension of these pairs and the segment

tension profile has now a symmetrically convex shape with an average lower tension

compared with the intermediate steps (black lines in Figure 3C). The same process

repeats itself in the formation of the subsequent segments.

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Figure 3 - Tissue segmentation from wave of apical constriction
(**A**) Time-series of a simulation showing the sequential segmentation of a tissue due to a linear increase

of apical cytoskeletal activity as cells are progressively activated from anterior (right) to posterior (left).

Colors as in Figure 3A. Snapshots taken at the approximate moment the clefts are formed (17000, 21000,

24000, 28000 and 31000 MCS). (**B**) Time evolution of cell pair tensions for cells in the regions of 2

consecutive segments. Cells are counted from the anterior position (right). (**C**) Tension profiles for the 2

tissue segments shown in (**B**). Each line corresponds to the same-color arrowheads in (**B**) and indicate

three different events - formation of the anterior segment boundary (green), formation of posterior

segment boundary (blue), and some time after the posterior event occurs (black).

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Segment size scaling with wave speed and rate of apical constriction

One of the attractive features of the clock and wavefront model, even before the

identification of the first oscillating gene in the PSM, was its ability to explain how somite

size can adjust to variations of embryo size: a faster clock would produce smaller

somites, while a faster wavefront would generate larger somites. We now investigate if

our mechanical model of segmentation has the same scaling features: does a faster

wave of activation (W) lead to larger segments? And how does this scaling change with

the associated build-up rate of apical contractility (Λ)? In the results that follow we

systematically varied both parameters and all data points have been averaged over 5

simulations with 115 to 205 cells.

We first fixed the build-up rate of apical contractility Λ and varied the wave speed W. We

observe two regimes of the average segment size $< S >$ with respect to the wave

speed. For wave speeds below a critical value ($< W^*$) average segment size $< S^* >$

was constant, but increases as a power law with exponents close to ¼ for faster wave

speeds ($> W^*$) (Figure 4A). The critical value of the wave speed W^* for the change to

the scaling regime, as well as the average segment sizes $< S^* >$ before scaling

depends on the value of the build-up rate Λ (Figure 4B). These results suggest that the

discrete spatial compartmentalization of the tissue in cell units poses a lower limit on

segment size as a function of Λ (Figure 4B). In fact, our simulations in absence of a

wave already indicates that the maximum tension a cell pair can reach depends on the

number of cells within a forming segment (Figure 2B), so for slow waves we expect the

cleft always appearing at the same size intervals.

Figure 4 - Segmentation as a function of wave speed
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(**A**) Average segment size $< S >$ as a function of wave speed W. For short wave speeds segments are of

roughly constant size, but scale as a power law afterwards. (**B**) Constant segment sizes $< S^* >$ (blue

solid line) and critical wave speeds W^* (red dashed line) as functions of Λ. (**C**) Spatial delay between

activation and boundary formation $< D >$ scales linearly with W. (**D**) Average segmentation time $< \tau >$

decreases as a power law with $W^{-0.80 \pm 0.008}$. (**A**,**C**,**D**) Each line shows a similar behavior for different base

values of the build-up rate of apical contractility (Λ).

The average distance between apical constriction activation and boundary formation $<$

$D>$ showed a less surprising behavior with an approximate linear scaling with wave

speed (Figure 4C): the faster the wave, the larger the distance between the wavefront

and the boundaries. Changes in the base value of Λ only shifted the curves, with lower

build-up rates increasing the spatial delay between activation and boundary formation.

The average time interval between successive boundary formation $<\tau>$, however, was

insensitive to the build-up rate of apical contractility Λ. This output only depends on W,

with faster wave speeds decreasing the segmentation time as a power law with

exponent -0.80 ± 0.008.

Figure 5 - Segmentation as a function of apical contractility build-up rate

(**A**) Average segment size $<S>$ as a function of build-up rate of apical contractility Λ. Segment size

decrease logarithmically with build-up rate, but stay constant for higher values of Λ. (**B**) Constant

segment sizes $<S^*>$ (blue solid line) and critical build-up rates Λ^* (red dashed line) as functions of W.

(**C**) Spatial delay between activation and boundary formation $<D>$ decreases linearly with Λ. (**D**)

Average segmentation time $<\tau>$ as a function of Λ. Each individual curve has the same shape as the

equivalent $<S>$ vs. Λ curves in (**A**). (**A**,**C**,**D**) Each line shows a qualitative similar behavior for different

base values of the wave speed (W).

Next we fixed the wave speed W and varied the build-up rate of apical contractility Λ.

Average segment sizes $<S>$ decrease logarithmically with higher build-up rates, but

became constant after a critical value of Λ^* (Figure 5A,B). As before, segment sizes

outside the scaling regime ($< S^* >$) depends on the value of the wave speed W used in

the simulations (Figure 5B). Again, these results can be understood in light of the way

cell pair tension scales with segment size (Figure 2B). For slow build-up rates ($< \Lambda^*$),

the sole factor determining segment size is the wave speed, with faster waves adding

more cells to the forming segment before the boundary formation (Figure 4A). For

higher build-up rates ($> \Lambda^*$), however, the rate of cell addition is not fast enough to

overcome the increase in tension profile, which itself is a function of tissue size (Figure

2B).

The spatial delay between cell activation and boundary formation $< D >$ decreased

linearly with the build-up rate, as expected: the faster the contractility build-up rate, the

less time it takes for segments to form once all cells are actively constricting (Figure

5C). The average segmentation time $< \tau >$ had the same dependency on the build-up

rate as the segment size $< S >$, as expected, given that $< S >= W < \tau >$ (Figure 5D).

Ratio of build-up rate to wave speed sets the transition of scaling regimes

The critical values of wave-speed W^* and build-up rate of apical contractility Λ^* for the

transition between constant and segment size scaling regimes shown in Figures 4B and

5B are in fact related. Rescaling of the horizontal axis in Figures 4A and 5A for each

curve by their corresponding values of Λ and W, respectively, shows that in both cases,

the transition occurs around the same ratio of $\Lambda/W = 22\ MCS^2/cell$ (Figure 6A,B). This

allows us to define a boundary transition in parameter space that separates regions

where segment sizes scales with variations in either Λ or W from regions where the

segment sizes remain relatively constant with changes in one of these parameters

(Figure 6C, green and blue regions). Note that we use the word *constant* in contrast to

the size scaling of the segments with respect to either Λ or W, there is still a small and

gradual change in segment sizes within the green region in Figure 6C (see

Supplemental Figure 1).

The parameter space in Figure 6C also contains a grey region where the combinations

of Λ and W leads to simulation artifacts. All data from this region were discarded in our

analysis. The orange region and dashed boundary line will be defined momentarily.

Figure 6 - Segmentation size scaling behavior
(**A**,**B**) Rescaled version of Figures 4A and 5B, showing average segment size $<S>$ as a function of (**A**)

wave speed to build-up rate, and (**B**) its inverse. Vertical dashed line at $\Lambda/W = 22$ indicate transition

threshold between constant size and scaling segment sizes. (**C**) Parameter space diagram showing

regions where the combination of the parameters Λ and W leads to constant segment sizes (green),

scaling segment sizes (blue), or irregularly sized segments. Black dashed line show transitions from

stable to splitting segments. Black X point indicates the reference simulation parameters (see Table 1).

Grey region indicate parameters combinations that lead to simulation artifacts.

Segment sizes become irregular for low ratios of build-up rate to wave speed

While average segment size keeps scaling with higher wave speeds (W) or lower build-

up rates of apical contractility (Λ), segment size distribution becomes irregular for higher

ratios of Λ/W (Figure 7A-D). This can be visualized in plots of segment size variation,

calculated as the ratio of standard deviation (σ) to the mean ($<S>$) (Figure 7E,F). This

happens because when the wave speed is much faster than the build-up rate ($W \gg \Lambda$),

a large group of cells starts to constrict at about the same time, the memory of the last

cleft position is dissipated, and new clefts are more likely to appear anywhere in the

tissue. We choose to classify segments as irregular when this variation is higher than

0.2. This corresponds, in our parameter space diagram, to a transition from regularly to

irregularly sized segments at build-up to wave ratios of $\Lambda/W \sim 4$. Segmentation time

variation ($\sigma/<\tau>$) is about 50% higher than size variation and had a similar

dependence with Λ/W (see Supplemental Figure 2). This higher variation in

segmentation time is expected from our mechanical model as the time of cleft formation

is an emergent result rather than a direct input as in the clock and wavefront model.

Figure 7 - Segment size variation
(**A**) Typical simulation output showing wide range of segment sizes from the irregular region (orange) of

the parameter space in Figure 6C. (**B**-**D**) Histogram of segment size distributions for combinations of

Λ and W in the constant size region (**B**), scaling region (**C**), and irregular region (**D**), of the parameter

space over 5 trials. (**E**) Segment size variation (std/mean) as a function of build-up rate of apical

cytoskeletal activity Λ. (**F**) Segment size variation as a function of wave speed W. (**E**,**F**) Dashed red lines

at $\sigma/<S> = 0.2$ indicate threshold used to draw determine the region of irregular segment sizes in Figure

6C.

Apical constriction sets a higher limit on segment sizes

The above results show that initial segment sizes increase with faster wave speeds

(Figure 4A) and slower build-up rates of apical contractility (Figure 5A). However, there

seems to be a higher limit on segment sizes. After their initial formation, larger

segments are prone to split as the cells within them continue to increase their apical

cytoskeletal activity strength (λ_A) and tension keeps building up until a new splitting

event occurs near the middle of the formed segment (Figure 8A,B).

While the occurrence of a splitting event is proportional to the initial segment size

(Figure 6C), the wave speed W has a more direct effect on the frequency of splitting,

with higher build-up rates only delaying the appearance of splits (Figure 8D). Together

these results suggest that our mechanical model of segmentation sets a higher limit on

segment sizes depending on the speed of the wave with respect to the build-up rate of

apical cytoskeletal activity (Figure 8C,D). Our results, however, did not point out to a

specific ratio of Λ/W as predictive of a transition to a higher frequency (>0.1) of splitting

events, but rather a nonlinear function $\Lambda = 0.6 * W^{0.7}$ for the boundary between stable

segments and splitting segments (dashed line in Figure 6C).

Figure 8 - Segment splitting
(**A**) Simulation snapshots showing the formation a splitting of a large segment (times, from top to bottom:

47000, 49000, 52000, 54000 and 56500 MCS). (**B**) Evolution of the tension profile for the splitting

segment shown in (**A**). (**C**) Plot of average final versus initial segment sizes. Data points at the diagonal

gray slashed lines indicate no splitting events. (**D**) Splitting frequency as a function of wave speed W.

Segment size as a function of maximum adhesion junction tension

So far, we had used a fixed apical to apical tension breaking threshold $T_{break} = -7500$.

We now test the effects of varying the threshold at which tension causes the coupling

between neighboring cell's apical domains to break. At the extreme limit of $|T_{break}|$ near

zero, the average segment size gets closer to 1 cell unit. Conversely, at high enough

values, the tissue never segments and remains a single monolayer. As the magnitude

of the breaking threshold increases between these ranges, the average segment sizes

also increase, as expected (Figure 9A). Note however, that the splitting frequency is

higher for low values of $|T_{break}|$, and zero for high values, with segments breaking in

more than once for threshold values around $T_{break} = -2000$ (Figure 9B).

Figure 9 - Influence of junctional tension limit and cell shape on segmentation
(**A**) Average segmentation size $< S >$ as a function of the absolute value of tension breaking point

$|T_{break}|$. Blue lines are initial segment sizes, red lines are segments sizes after splitting. (**B**) Average

number of splitting events per segment as a function of $|T_{break}|$. (**C**) Average segmentation size $< S >$

(blue) and spatial delay between activation and boundary formation $< D >$(red) as functions of cell aspect

ratio AR. (**D**) Average segmentation time $< \tau >$ (blue) and frequency of segmentation splitting (red) as

functions of cell aspect ratio AR.

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Segmentation as a function of cell shape aspect ratio

In the somitic tissue, cells dynamically elongate while the process of boundary formation

takes place. Since we have kept the length of the cell fixed for our simulations, we now

evaluate how segment size depends on the aspect ratio of the cell ($AR =$

$height/width$). Figure 9C shows the average segment size $< S >$ and delay $< D >$ as a

function of cell aspect ratio. As expected, the more elongated they are the more cells

can be packed into a segment. The delay, however, is a concave curve with respect to

AR with a maximum around $AR = 2$. This can be understood as follows: the wider the

cells ($AR < 2$) the harder it is to pack them into a segment made of wedge-shaped cells,

so apical tension builds up quickly and clefts happen sooner, thus reducing both the

segmentation time (blue line in Figure 9D) and the spatial delay between activation and

boundary formation compared to the standard case (Figure 9C). Conversely, the more

elongated cells ($AR > 2$), the greater the number of cells that can be packed into a

segment and the longer it takes for consecutive boundaries to form. However, since the

segments themselves are larger, the spatial delay, which is calculated in terms of

segment sizes, also decreases. Figure 9D also shows that the segment splitting

frequency (red line) is significantly higher for low aspect ratios, which can be understood

as a result from the packing constraints that expedites the accumulation of tension

between apical cell pairs and make them more prone to splitting.

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DISCUSSION

In this paper, we developed a mechanical model of tissue segmentation where a

continuous wave of apical constriction activation coupled with maximum apical tension

load set future somite boundaries. Our modeling effort is based on experimental

observations of intensive cell rearrangements during somite formation (Martins et al.

2010) and our own SEM images (Figure 1A) that suggests that an early segmentation

process is occuring at the dorsal side of the PSM about 4 somite lengths prior to the

formation of the last somite. We interpreted the spatial segmentation of the dorsal

epithelium into similarly sized cohorts of cells as the result of an increased longitudinal

(dorsal-caudal) tension between the cells. A similar process is known to cause the

appearance of periodic cracks on non-biological thin films subjected to tensile stress

(Thouless (1990), Thouless et al (2011)). Here, however, our material is subdivided into

discrete units; cannot be assumed as purely elastic; and the source of stress does not

come from outside but is: i) generated internally by the cells, ii) is time-varying; and iii) is

spatially inhomogeneous at any point of time. Given those differences we developed our

own model of epithelial segmentation where we identified four key parameters: the

speed of the wave (W), the rate of increase of apical contractility (Λ), the maximum

apical cell-cell junction load (T_{break}), and the aspect ratio of the cells (AR) (Table 1).

While the parameters have been chosen to imitate segment size in chicken, we

comment that we need better tuning of the aspect ratio of the cells in the dorsal region.

By the time the somites form, cells in this region can reach aspect ratios as high as 7.

To compare results at these higher aspect ratios we would require re-parametrizing our

cell volume to allow for sufficient apical surface so that the width is much greater than a

chosen L_{AT}. While these changes are easily programmable, they come at a greater

computational cost.

We found that our purely mechanical model is able to produce spatially and temporally

regular segment sizes. Similar to the CW models, segment sizes scale (increases) with

higher wave speeds. (Figure 4A). We found, however, that this scaling is not linear with

wave speeds, as would be expected from a simple deterministic version of the CW

model. There are two limits to this scaling behaviour, which can be characterised by

Λ/W ratios alone (Figure 6C). An upper limit in segment sizes is reached for low ratios

($\Lambda/W < 4$), where initially formed segments are prone to splits due to mechanical

instabilities; and a lower limit in the sizes is reached for high ratios ($\Lambda/W > 22$), where

segment sizes do not scale with wave speeds. These limits in somite size are absent in

CW models. Segmentation time in our model is not imposed but rather calculated as the

difference in time for the appearance of the rostral and caudal boundaries. We find that

it decreases with wave speeds and is insensitive to build-up rates (Figures 4D, 5D).

Ou model seems to have higher variations in the distribution of the segment sizes and

formation time than existing mathematical implementations of the clock-and-wavefront

model. In a real 3D scenario however, epithelialization along the dorsal sides does not

happen along a line as in our model, but within plane that also extends along the

medial-lateral direction. Addition of these neighboring cells into our model would likely

reduce variation in segment size and segmentation time. It is important to note that *in*

vivo this early segmentation will be succeeded by epithelialization along other

directions/planes reinforcing the initial boundaries and reducing the noise in the

segment size distribution.

While our mechanical model drastically differs from the CW family of models with

respect to the absence of a somitic intracellular clock, it still assumes the presence of

some sort of caudally travelling wave, which does not need to be the same as the one

postulated by the CW. We are agnostic about the nature of this wave, which can be

either a cell-autonomous maturation process (starting from their addition to the tail-bud

or a similar event), a tissue level processes, such as a read-out of FGF and RA levels,

or a combination of both.

Finally, we are not proposing an alternative model for somitogenesis that replaces the

proposed CW model, but rather exploring a complementary mechanism that is likely

taking place in addition to an intra- and inter-cellular signalling mechanism. We found

that our simple mechanical model of tissue segmentation is able to recapitulate most

aspects of somitogenesis that a CW explain, such as the spatially and temporally

periodic formation of somites and the correlation between somite size and wave-speeds

as seen across species. In addition to that, our mechanical model also sets limits to the

minimum and maximum sizes of somites and suggest an explanation for the observed

splitting of large somites formed in some perturbation experiments (*Stern & Bellairs*

(1984), Cooke (1978)). As is often the case in biology, we speculate that the two

processes are likely operating in conjunction to yield a robust morphological outcome.

The combination of a mechanical and signalling process can explain, for example, the

jumping of cells between forming somitic boundaries (*Kulesa et al (2002)*) where a cell

mechanically constrained to be in one somite later moves and ingresses into the

neighboring somite in as a response to its CW read-out.

Our current model, however, is to simple to explore such scenarios as it is restricted to

a 1D view of the dorsal monolayer and is aimed to reproduce segmentation events prior

to the formation of a full somite. Further developments of this basic model will include

the self-organization of the PSM cells ventral to the dorsal monolayer and the expansion

of the model to 3D. It is known that in chicken (*Martins et al (2010)*) different regions of

the PSM epithelialize at different times, and it would be interesting to explore the

implications of these observations in an augmented version of the mechanical model

presented here.

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