

Three Dimensional Multi-Mode Rayleigh-Taylor and Richtmyer-Meshkov Instabilities at All Density Ratios

**D. Kartoon^{1,2}, D. Oron³, L. Arazi⁴, A. Rikanati^{1,2},
O. Sadot^{1,5}, A. Yosef-Hai^{1,5}, U. Alon³, G. Ben-Dor⁵
and D. Shvarts^{1,2,5}**

1. *Dept. of Physics, Nuclear Research Center Negev, Israel*
2. *Dept. of Physics, Ben-Gurion University, Beer-Sheva, Israel*
3. *Fac. of Physics, The Weizmann Institute of Science, Rehovot, Israel*
4. *School of Physics and Astronomy, Tel-Aviv University, Tel-Aviv, Israel*
5. *Dept. of Mech. Eng., Ben-Gurion University, Beer-Sheva, Israel*

Relevant Publications

Oron et al. PoP 2001.

Alon et al, PRL 1994, 1995.

Hecht et al. PoF 1994.

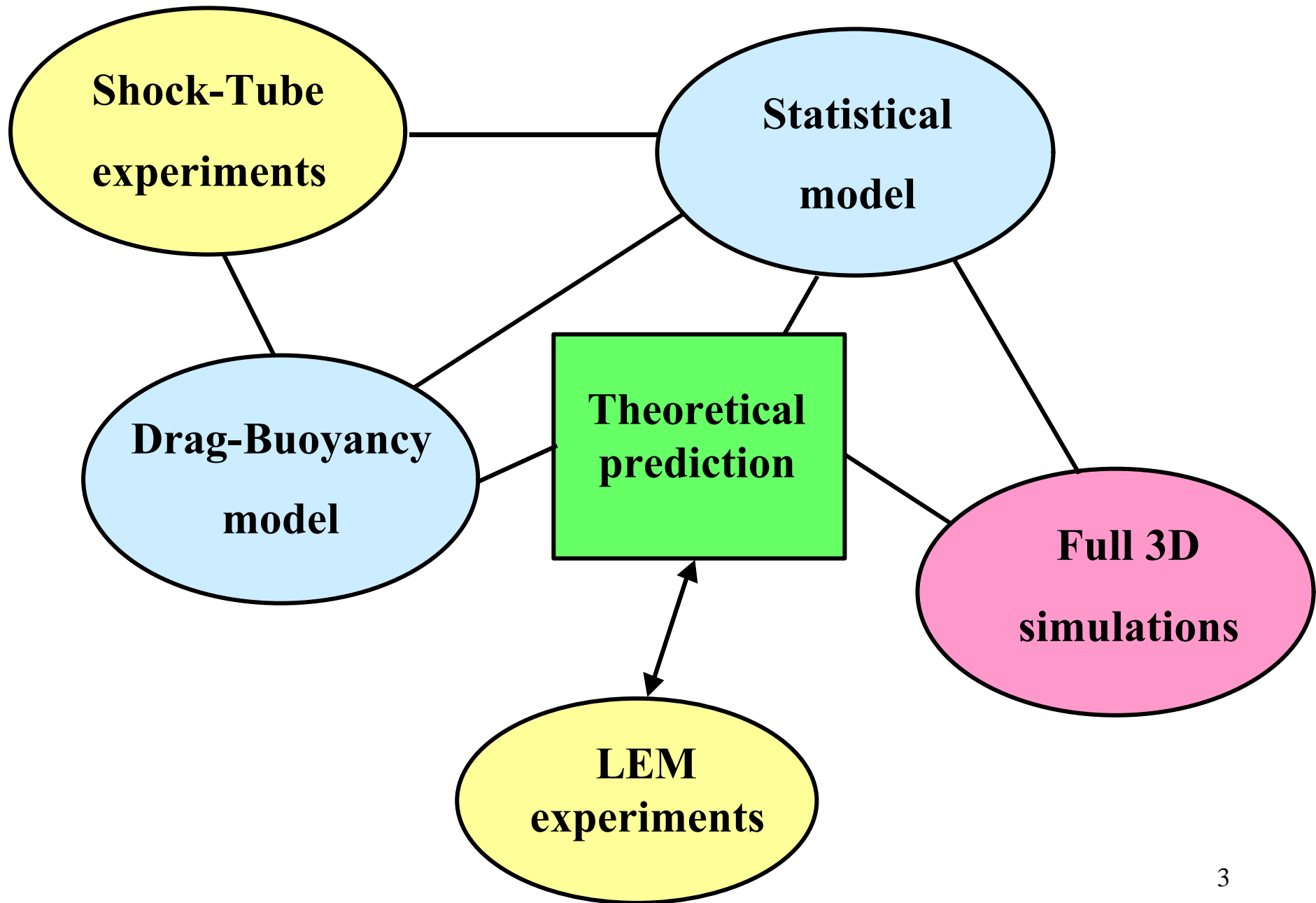
Oron et al. PoP 1998.

Rikanati et al. PRE 1998.

Sadot et al. PRL 1998.

Layzer APJ 1955.

Dimonte PoF, PoP 2000.



Outline

The Drag-Buoyancy Model for Bubbles

2D Statistical Model

Full 3D Numerical Simulations

3D Statistical Model

The Drag-Buoyancy Model for Spikes

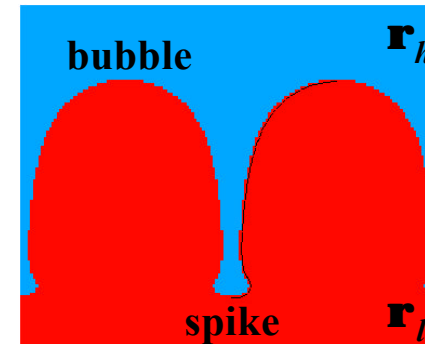
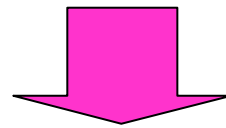
Conclusion

Single Mode Nonlinear Stage

$$(\mathbf{r}_l \cdot V + C_a \cdot \mathbf{r}_h \cdot V) \cdot \frac{du}{dt} = (\mathbf{r}_h - \mathbf{r}_l) V \cdot \mathbf{g} - \mathbf{r}_h \cdot S \cdot u^2$$

(inertia) + (added mass) = (buoyancy) (kinematic drag)

Where $V/S \propto \lambda$



$$(\mathbf{r}_l + C_a \mathbf{r}_h) \dot{u} = (\mathbf{r}_h - \mathbf{r}_l) \cdot \mathbf{g} - \frac{C_d}{I} \mathbf{r}_h \cdot u^2$$

Where C_a and C_d are geometric constants.

Single Mode Asymptotic Velocities

$$(\mathbf{r}_l + C_a \mathbf{r}_h) \dot{u} = (\mathbf{r}_h - \mathbf{r}_l) \cdot \mathbf{g} - \frac{C_d}{I} \mathbf{r}_h \cdot \mathbf{u}^2$$

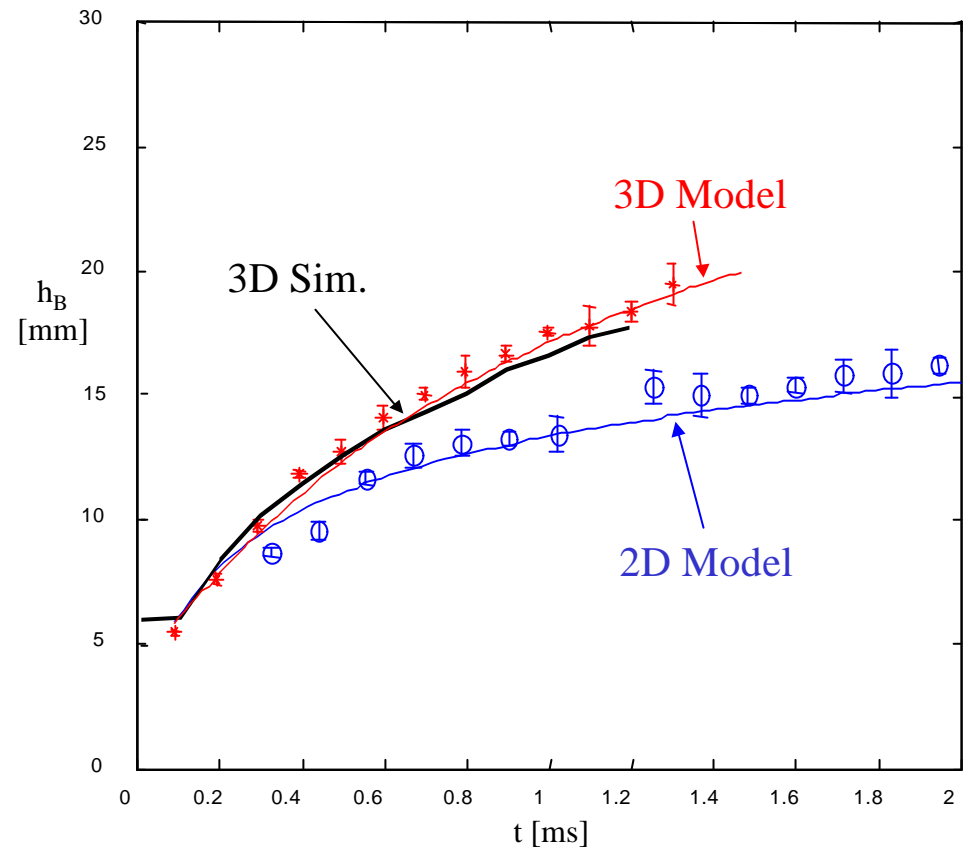
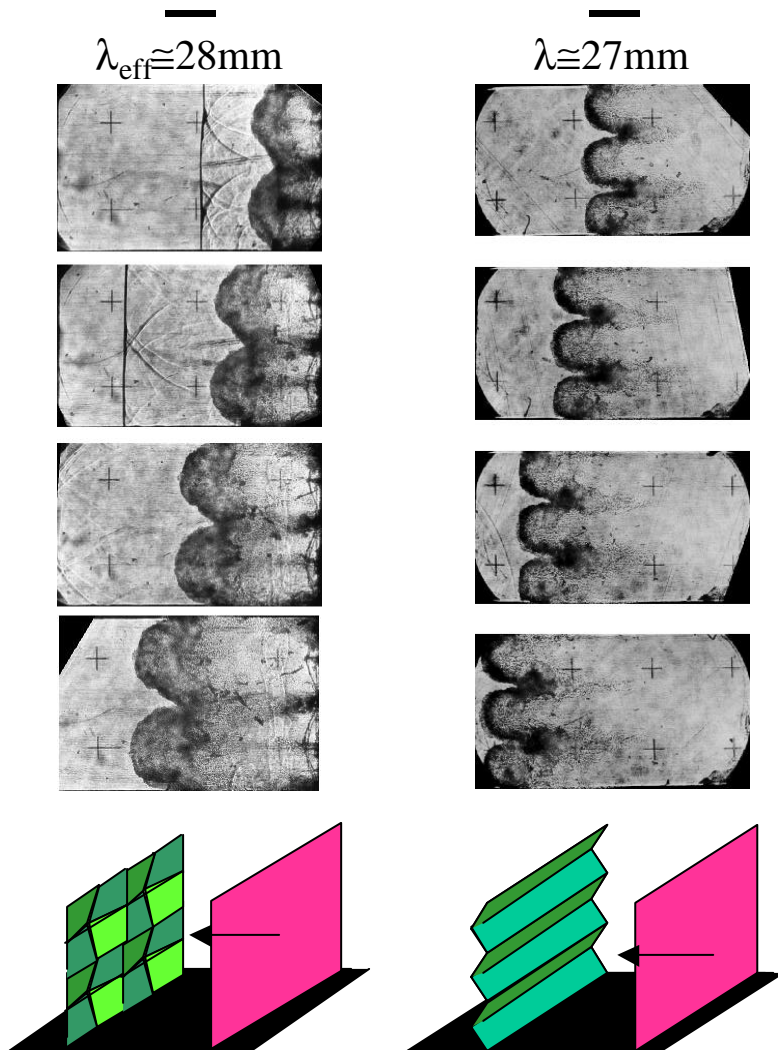
RT:
$$u_B = \sqrt{\frac{1}{C_d} \left(\frac{2A}{1+A} \right) g \lambda}$$

RM:
$$u_B = \frac{1}{C_d} \cdot \left(\frac{1-A}{1+A} + C_a \right) \cdot \frac{\lambda}{t}$$

The geometric constants depend on the dimensionality:

$$\begin{array}{ll} C_a^{2D} = 2, & C_d^{2D} = 6\pi \\ C_a^{3D} = 1, & C_d^{3D} = 2\pi \end{array}$$

RM Single Mode Experimental Results



Multi Mode Drag Buoyancy Model

Generalization of the drag-buoyancy equation, using the self-similarity assumption:

$$h_B^{(MM)}(t) = b(A) \cdot \langle \mathbf{1} \rangle(t)$$

RT: $u = \dot{h} = \sqrt{c_1(A) \cdot g \cdot \frac{h}{b(A)}} = \mathbf{a} \cdot A \cdot gt^2$

RM: $u = \dot{h} = c_2(A) \cdot \frac{h}{b(A)} \cdot \frac{1}{t} = c \cdot t^q$

$$\alpha = \frac{1}{2(1+A)} \frac{1}{C_d} \frac{1}{b^{\text{RT}}(A)}$$

$$\theta = \left(\frac{1-A}{1+A} + C_a \right) \cdot \frac{1}{C_d} \cdot \frac{1}{b^{\text{RM}}(A)}$$

3D Multi Mode the a q b Relations

Using the 3D coefficients in the expressions relating α , θ and b , and assuming that $b^{(RM)}=b^{(RT)}$, the differences between 2D and 3D bubble front growth are obtained:

	2D ($C_a=2, C_d=6\pi$)	3D ($C_a=1, C_d=2\pi$)
assumed a	0.05	0.05
q	$0.1(A+3)$ (=0.3-0.4)	0.2
b	$\frac{1}{2(A+1)}$ (=0.25-0.5)	$\frac{3}{2(A+1)}$ (=0.75-1.5)

Outline

The Drag-Buoyancy Model for Bubbles

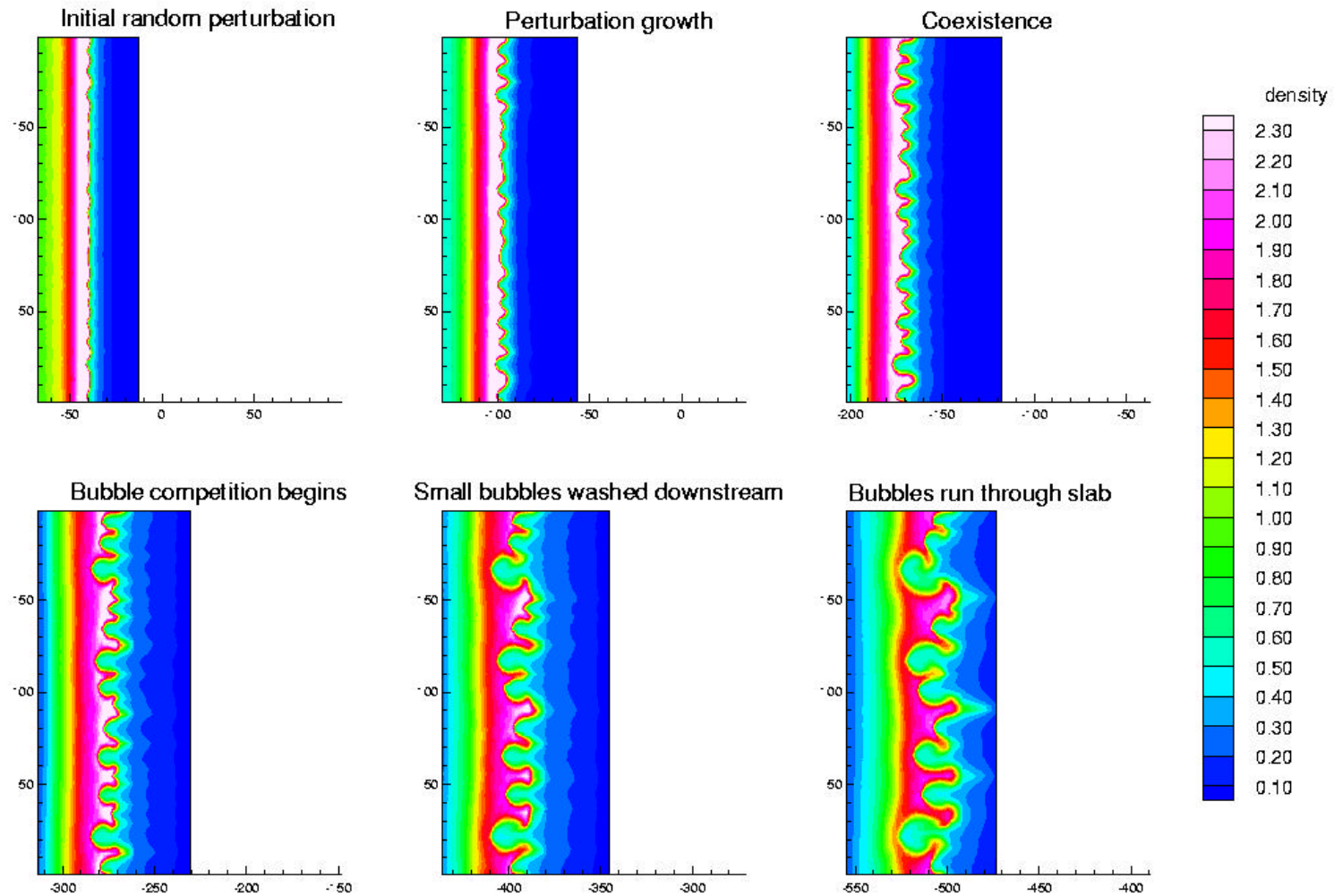
2D Statistical Model

Full 3D Numerical Simulations

3D Statistical Model

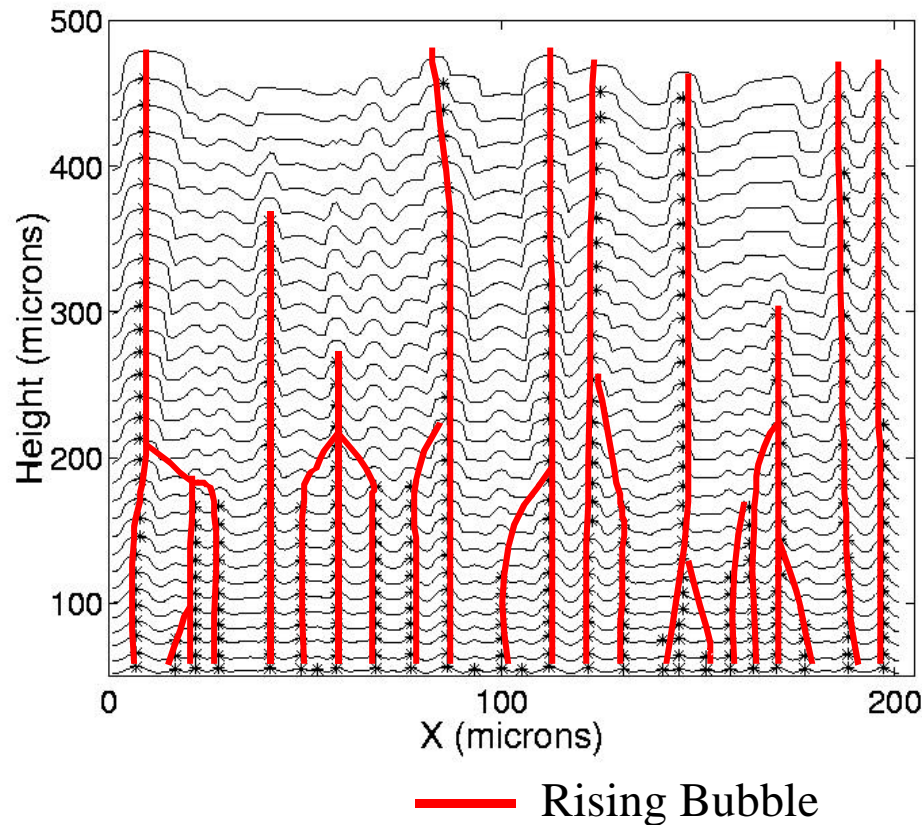
The Drag-Buoyancy Model for Spikes

Simulation of 2D Multimode Perturbation Evolution



Multi Mode 2D Statistical Model (Alon et. Al.)

Bubble Envelope (Simulation)

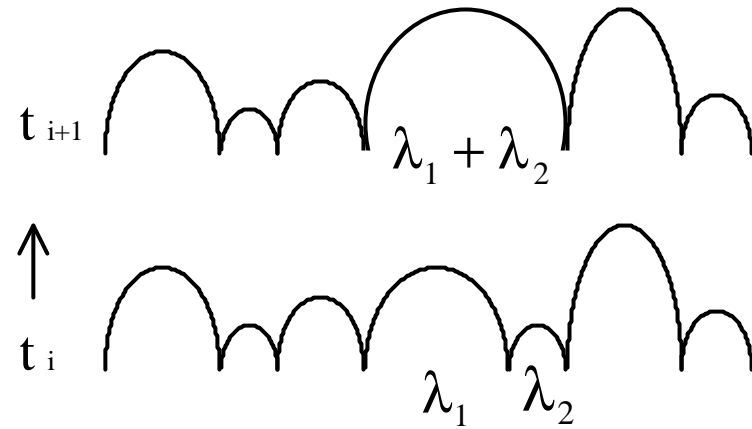
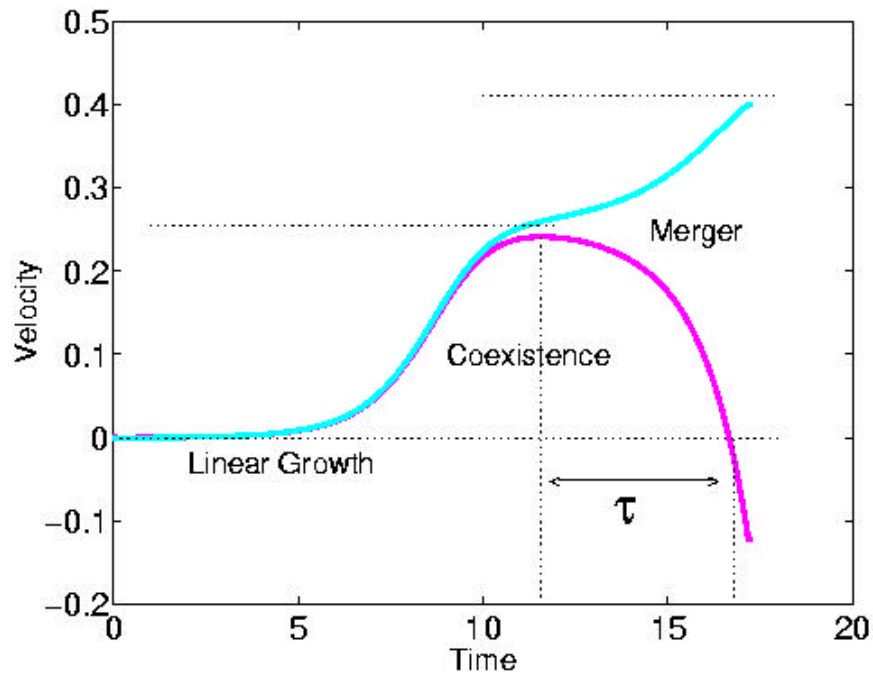


Each bubble grows with its asymptotic velocity, according to its wavelength:

$$\mathbf{u}_i = \mathbf{u}_{2D}^{(asy)}(\mathbf{I}_i, t)$$

Multi Mode Bubble Merger

Velocity evolution of two non-identical adjacent bubbles:



Merger Rate:

$$(\mathbf{l}_i, \mathbf{l}_{i+1})$$

$$\downarrow \mathbf{w}_{merge}(\mathbf{l}_i, \mathbf{l}_{i+1}, t) \sim 1/t$$

$$(\mathbf{l}_i + \mathbf{l}_{i+1})$$

Multi Mode 2D Statistical Model

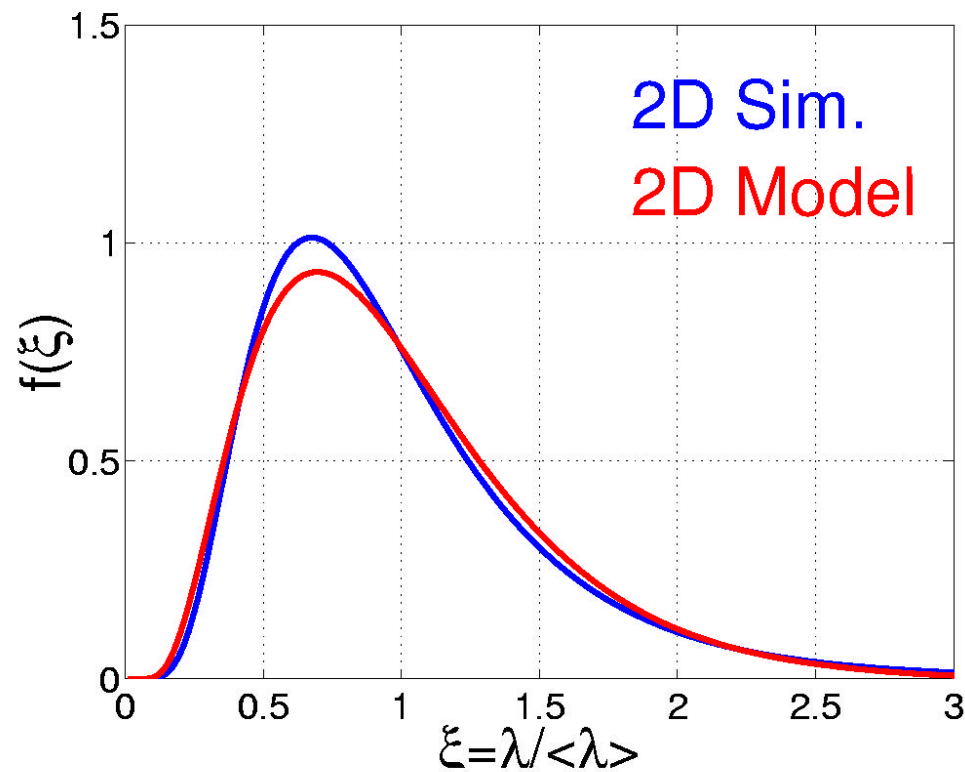
$$N(t) \cdot \frac{\mathcal{I}g(\mathbf{l}, t)}{\mathcal{I}t} = -2g(\mathbf{l}, t) \int_0^{\infty} g(\mathbf{l}', t) \mathbf{w}(\mathbf{l}, \mathbf{l}') d\mathbf{l}' \quad \text{death}$$
$$+ \int_0^{\infty} g(\mathbf{l} - \mathbf{l}', t) g(\mathbf{l}', t) \mathbf{w}(\mathbf{l} - \mathbf{l}', \mathbf{l}') d\mathbf{l}' \quad \text{birth}$$

Where $g(\lambda, t)$ is the number of bubbles with wavelength λ within interval $d\lambda$ at time t , and $N(t)$ is the total number of bubbles:

$$\frac{\partial N(t)}{\partial t} = -\langle \mathbf{w} \rangle_g \cdot N(t)$$

Multi Mode 2D Statistical Model Results

1. The λ distribution reaches an asymptotic function:



Multi Mode 2D Statistical Model Results

2. The average bubble and spike heights are obtained for both the RT and the RM case:

$$\text{RT: } u_B^{\text{asy}} \propto \sqrt{g\lambda} \xrightarrow{\text{self-similarity}} \dot{h}_B \propto \sqrt{gh} \rightarrow h_B = \alpha_B \cdot gt^2$$

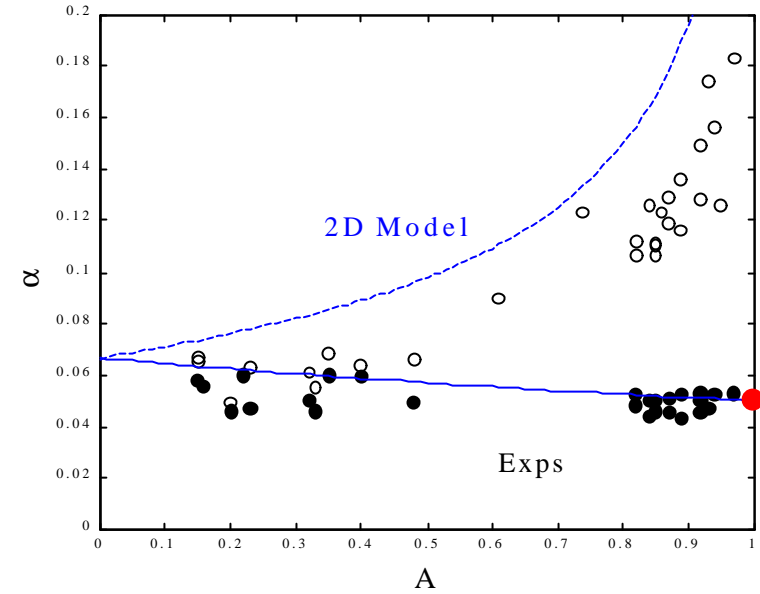
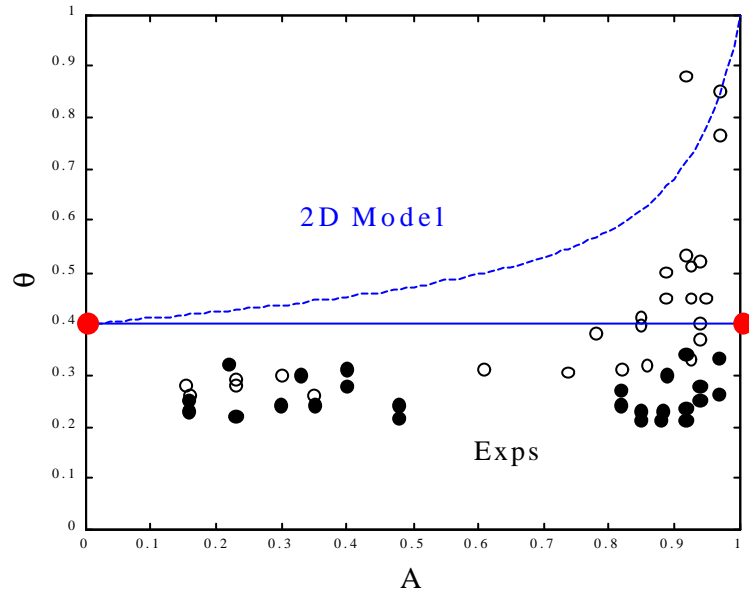
$$\text{RM: } u_B^{\text{asy}} \propto \frac{\lambda}{t} \xrightarrow{\text{self-similarity}} \dot{h}_B \propto \frac{h_B}{t} \rightarrow h_B = c \cdot t^{\theta_B}$$

$$\text{Self-similarity parameter: } b^{\text{RT}} = b^{\text{RM}} = \frac{h_B}{\langle \lambda \rangle}$$

2D Statistical Model Results:

$$\alpha_B \cong 0.05 \quad \theta_B \cong 0.4 \quad b \cong 0.2$$

LEM Experimental Results

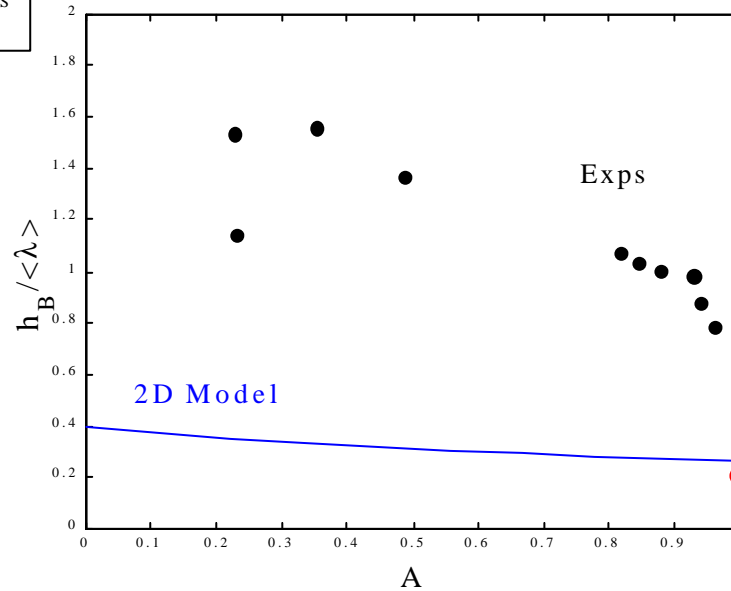


RM: $h_{B/S} = a_0 \cdot t^{\theta_{B/S}}$

RT: $h_{B/S} = \alpha_{B/S} \cdot A g t^2$

Drag-Buoyancy Model
Effective 1-Mode

2D Statistical Model



Outline

The Drag-Buoyancy Model for Bubbles

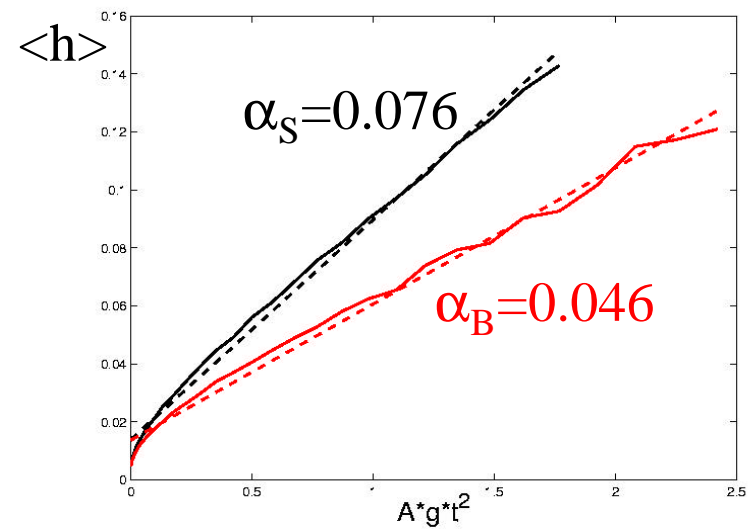
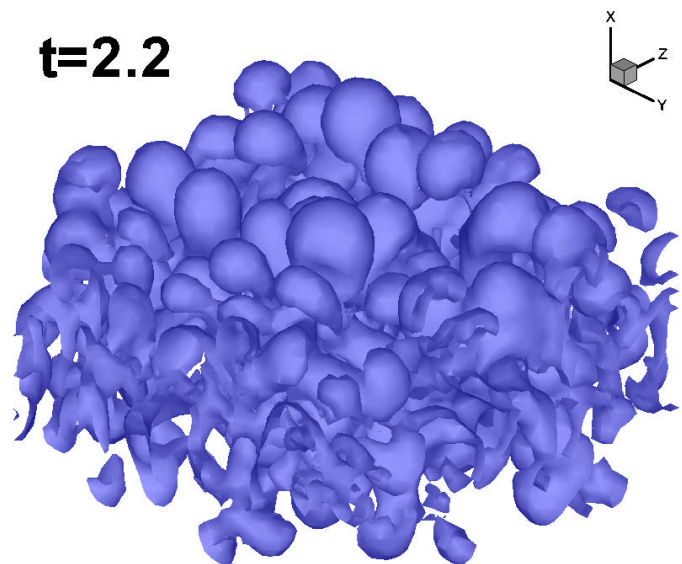
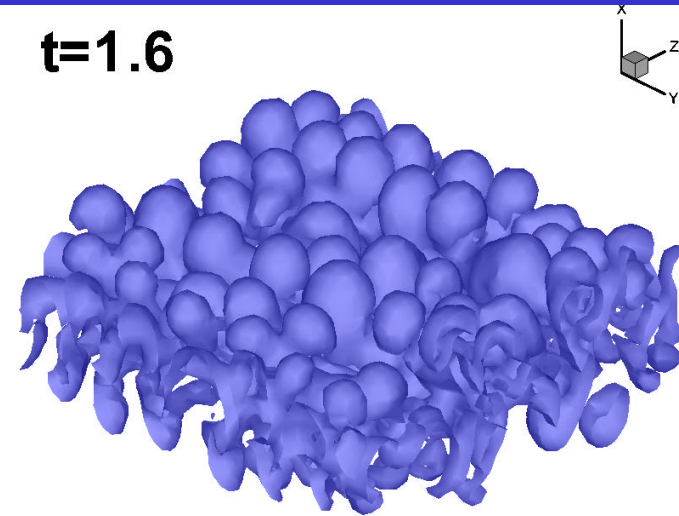
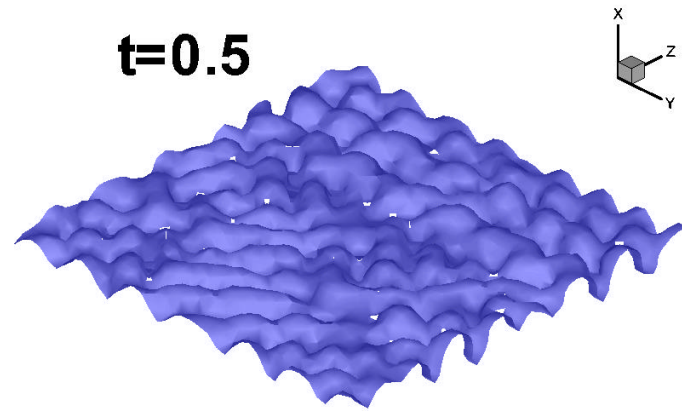
2D Statistical Model

Full 3D Numerical Simulations

3D Statistical Model

The Drag-Buoyancy Model for Spikes

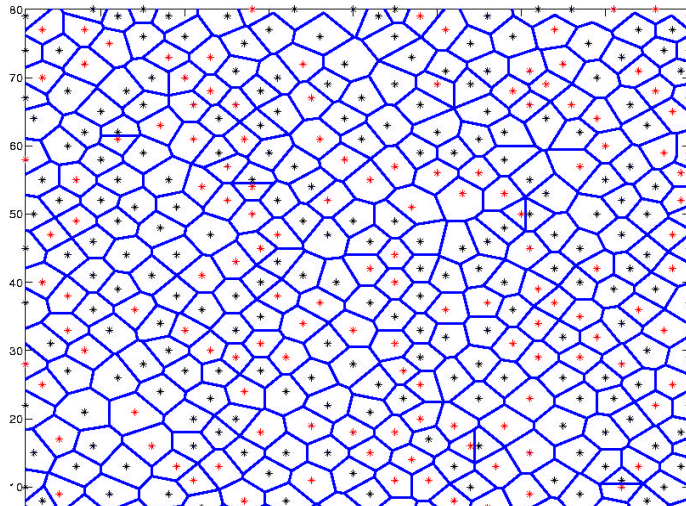
Full 3D Numerical Simulation of the RT $A=0.5$ Case



Voronoi Cell Structure of the Bubble Front Demonstrates the 3D Bubble Merger

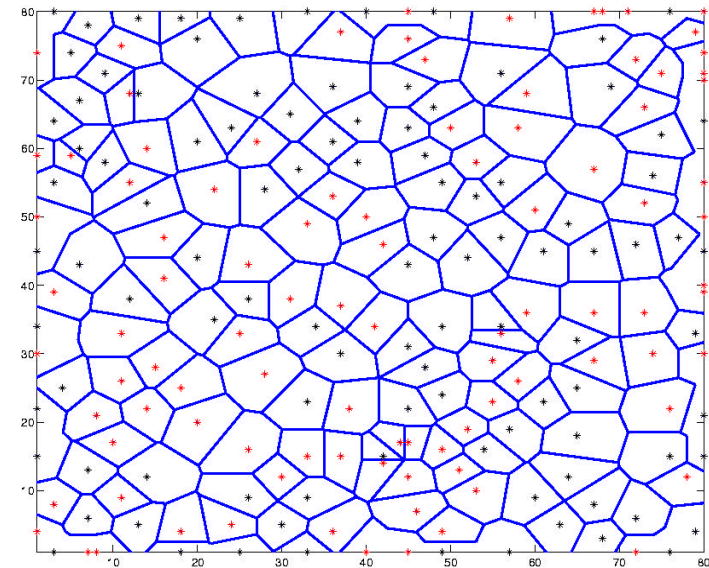
$t=0.28$

$n=260$



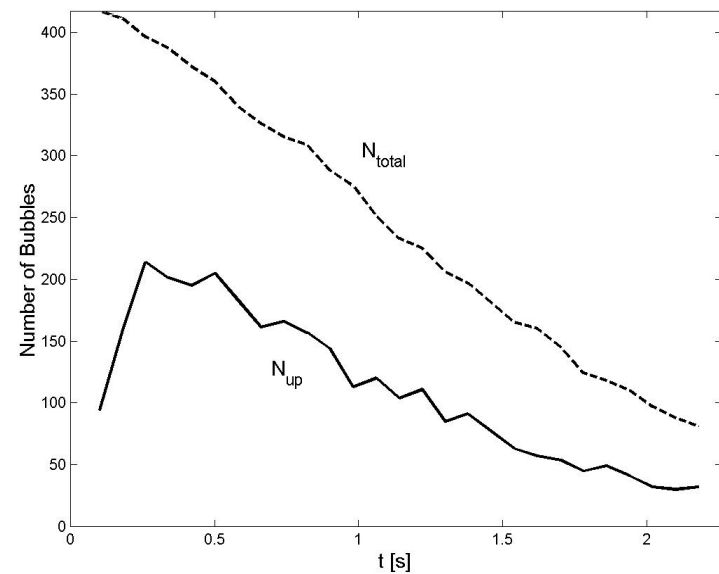
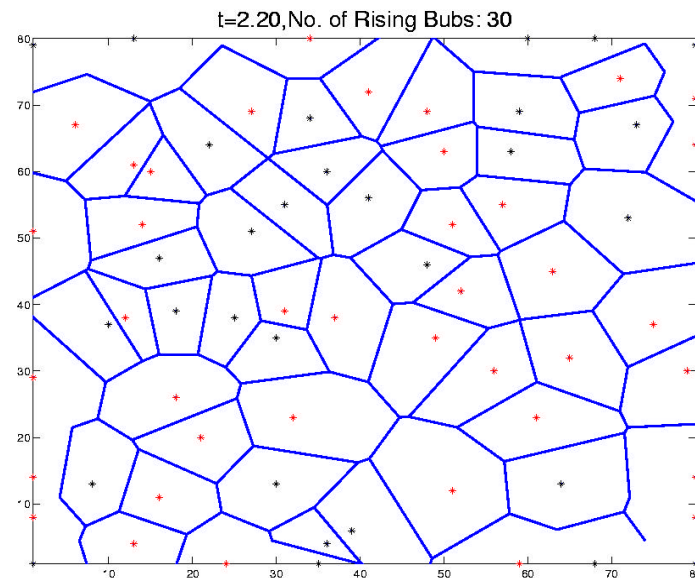
$t=1.4$

$n=112$

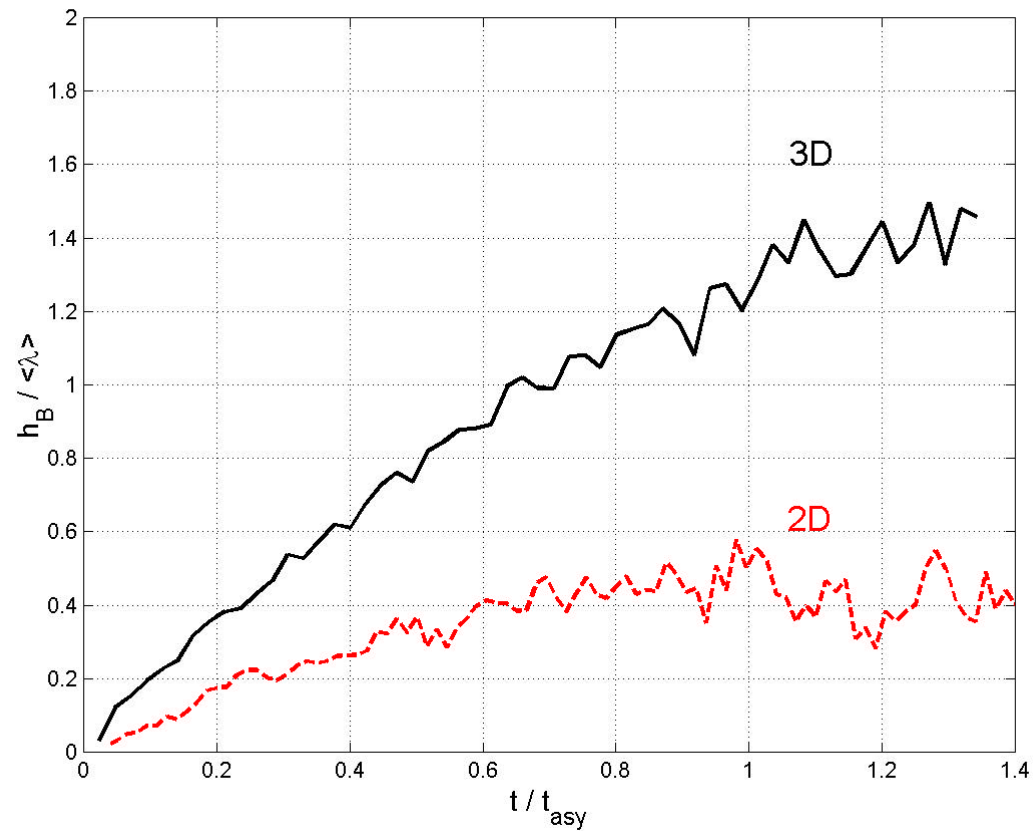


$t=2.2$

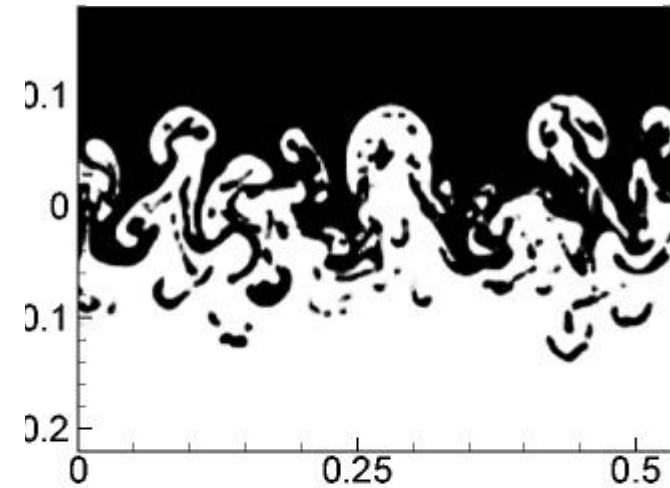
$n=30$



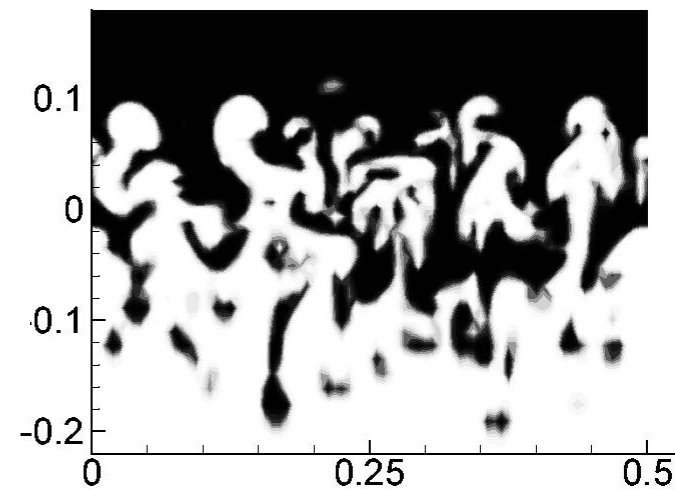
3D Simulation Results



2D Simulation



3D Simulation slice



Outline

The Drag-Buoyancy Model for Bubbles

2D Statistical Model

Full 3D Numerical Simulations

3D Statistical Model

The Drag-Buoyancy Model for Spikes

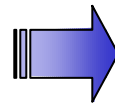
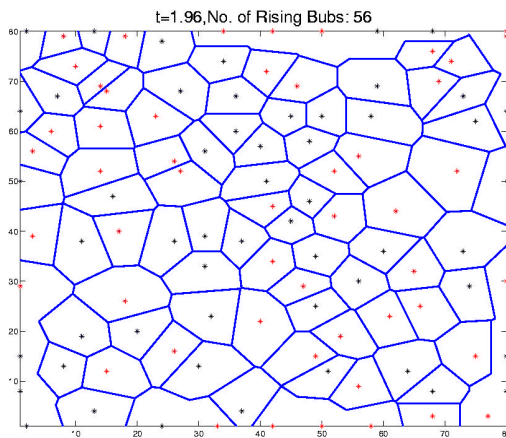
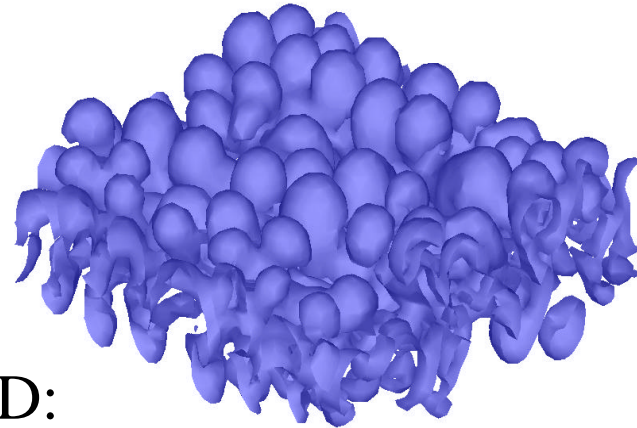
Multi Mode 3D Statistical Model

Dimensionality effects on the statistical model:

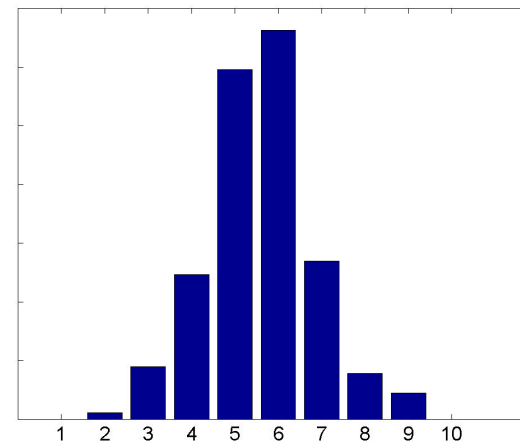
Asymptotic velocity of each bubble 1.5-2 times higher than in the 2D case:

$$u_{3D} = 1.5-2u_{2D}$$

Average number of neighbors per bubbles ≈ 6 in 3D, rather than 2 in 2D:



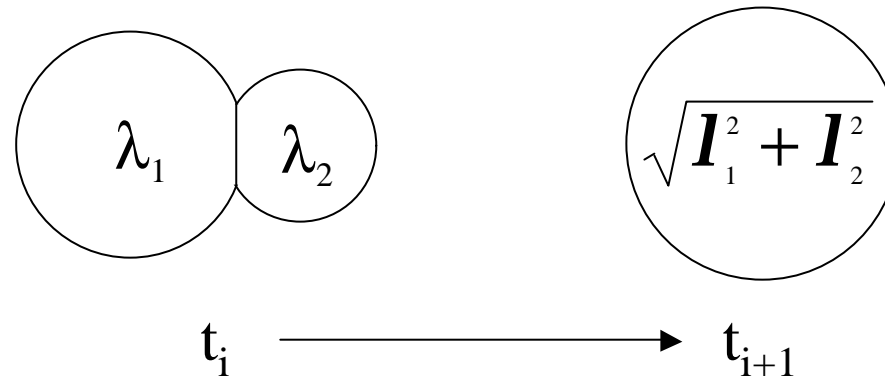
No. of Neighbors Distribution



Multi Mode 3D Statistical Model

Dimensionality effects on the bubble merger:

Bubble merging in 3D conserves area, rather than length in 2D:

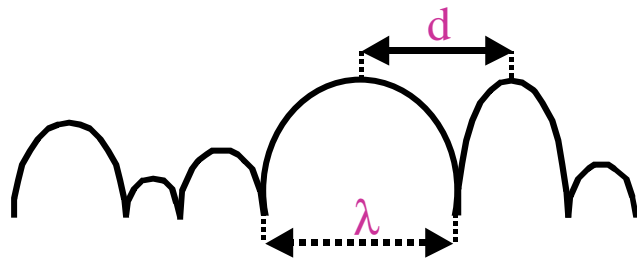


The merges occur with rate $\omega(\lambda_1, \lambda_2)$. At first step ω^{3D} was taken to be equal to $(u_{3D}^{asy} / u_{2D}^{asy}) \cdot \omega^{2D}$

Because of the area conservation, a 3D bubble has to merge with more of its neighbors in order to reach the same \mathbf{l} . This effect reduces $d\mathbf{l}/dt$, which in turn reduces both \mathbf{a} and \mathbf{q} (and increases \mathbf{b}).

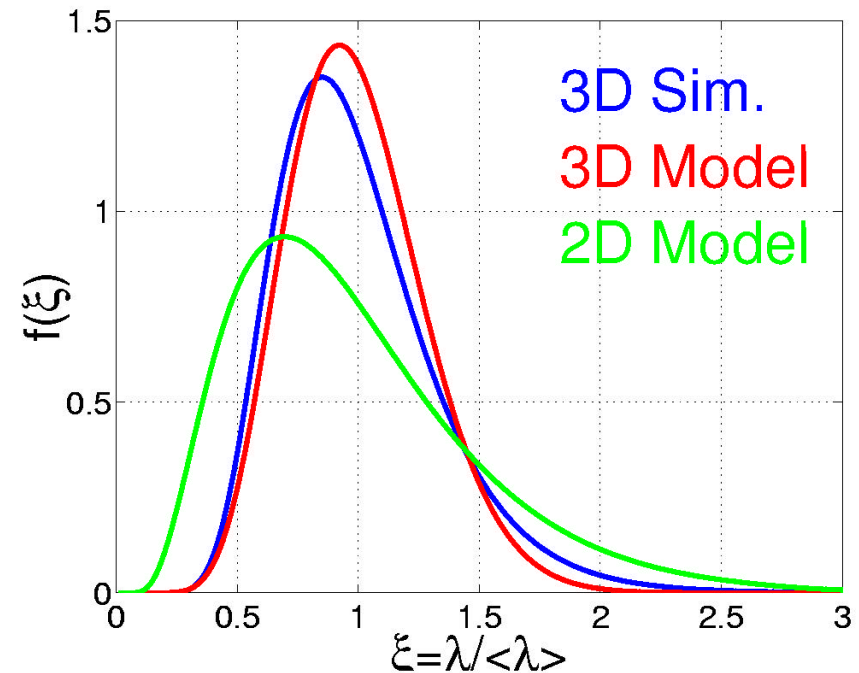
Multi Mode 3D Statistical Model Results

The segment distribution $g(d)$ is obtained from the simulation:



The relation between $g(d)$ and $g(\lambda)$ is given by:

$$g(d) = 2 \int_0^d g(2d - 2l)g(2l)dl$$



Multi Mode 3D Statistical Model Results

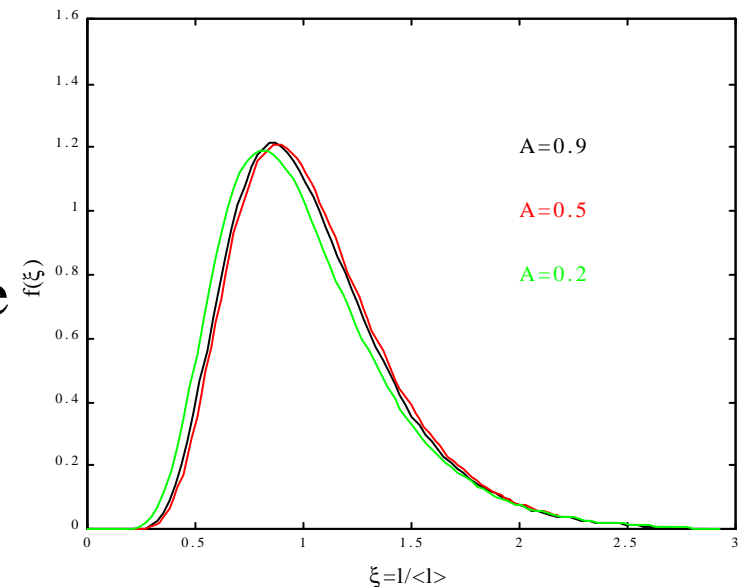
3D Statistical model results agree well with 3D Simulations.

$$\alpha_B = 0.055 \quad \theta_B = 0.18 \quad b = 0.67$$

The 3D wavelength distribution is narrower than the 2D distribution. The narrowing of the λ distribution is due to:

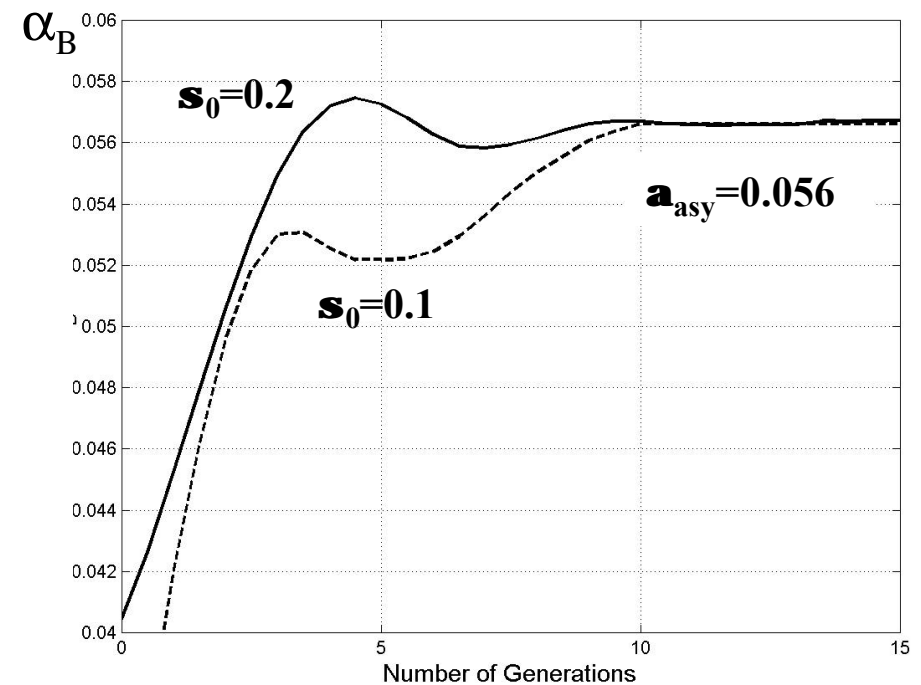
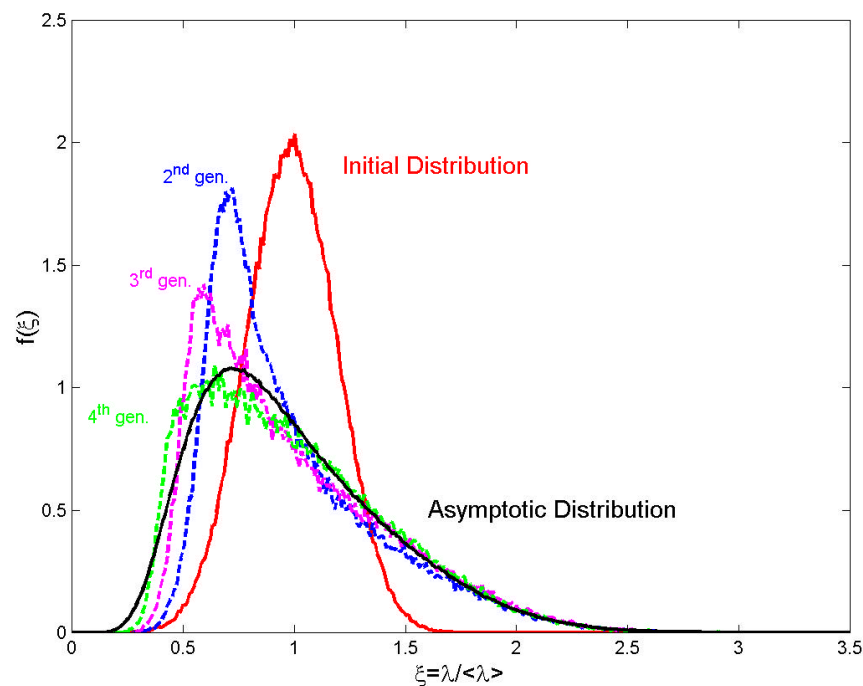
- Reduction of $d\lambda/dt$.
- Increased number of neighbors.

Simulations results indicate that the 3D statistical model may be applicable to a wide range of A:

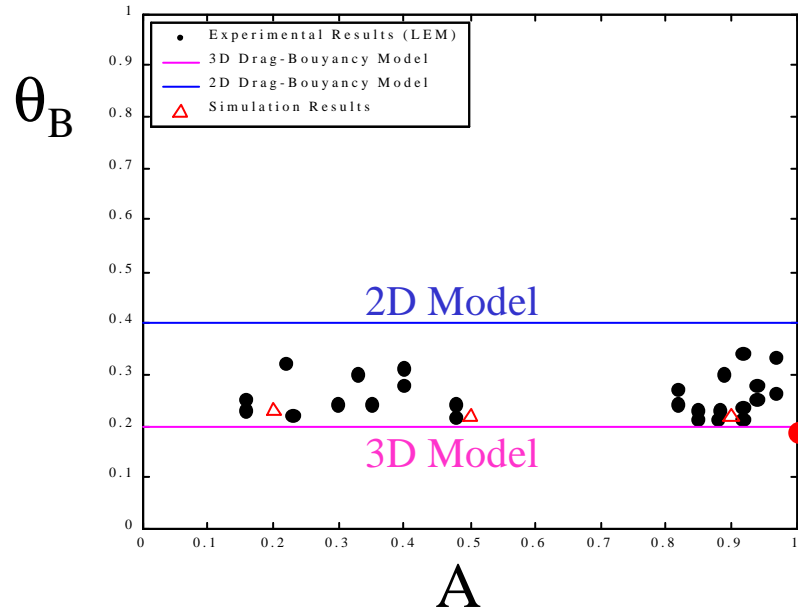
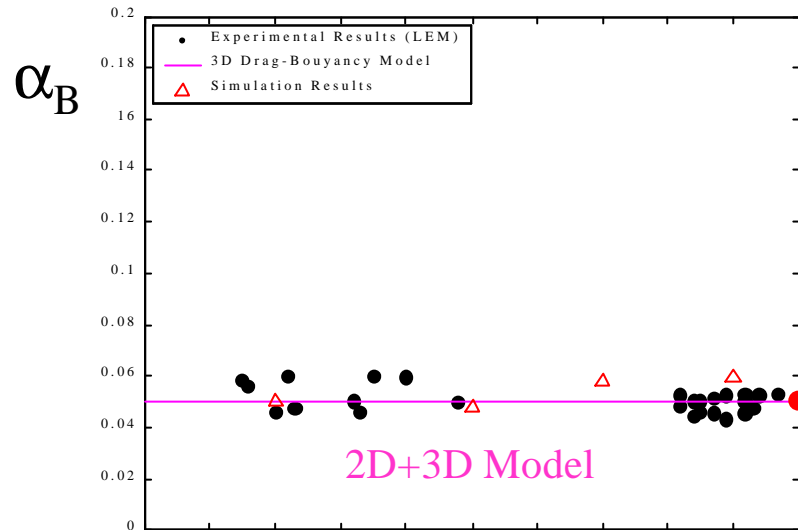


3D Statistical Model Dependence on Initial Distribution

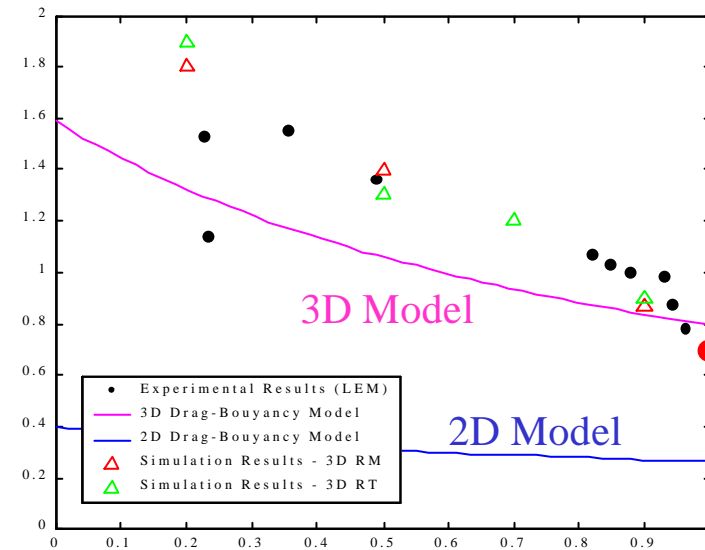
Using the initial wavelength distribution derived from the voronoi diagram in the statistical model gives the α dependence on the generation number:



LEM Experimental Results Vs. 3D Models and Simulations



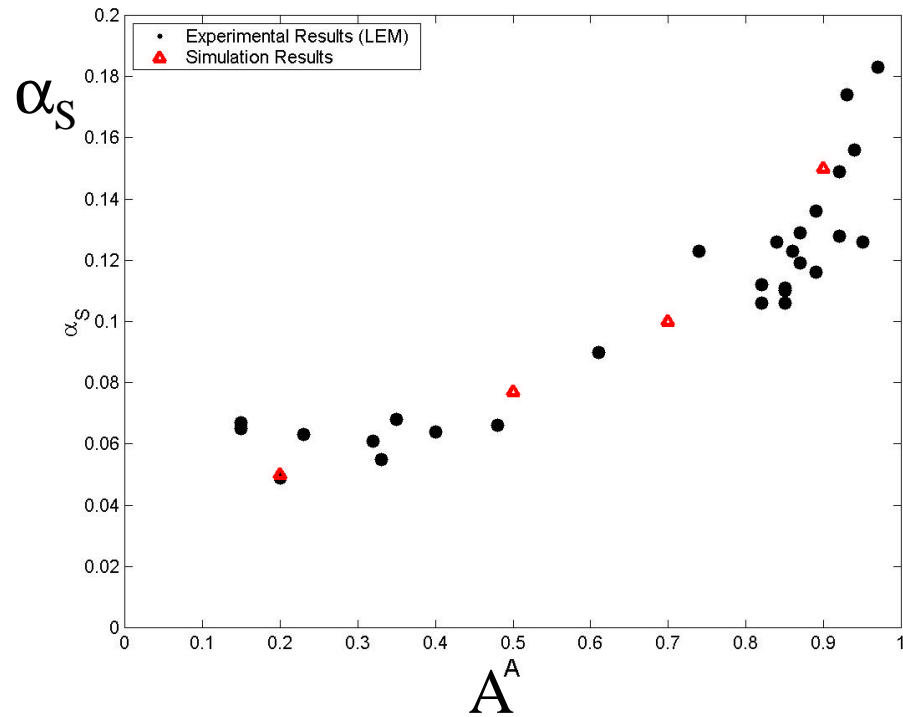
$h_B / \langle \lambda \rangle$



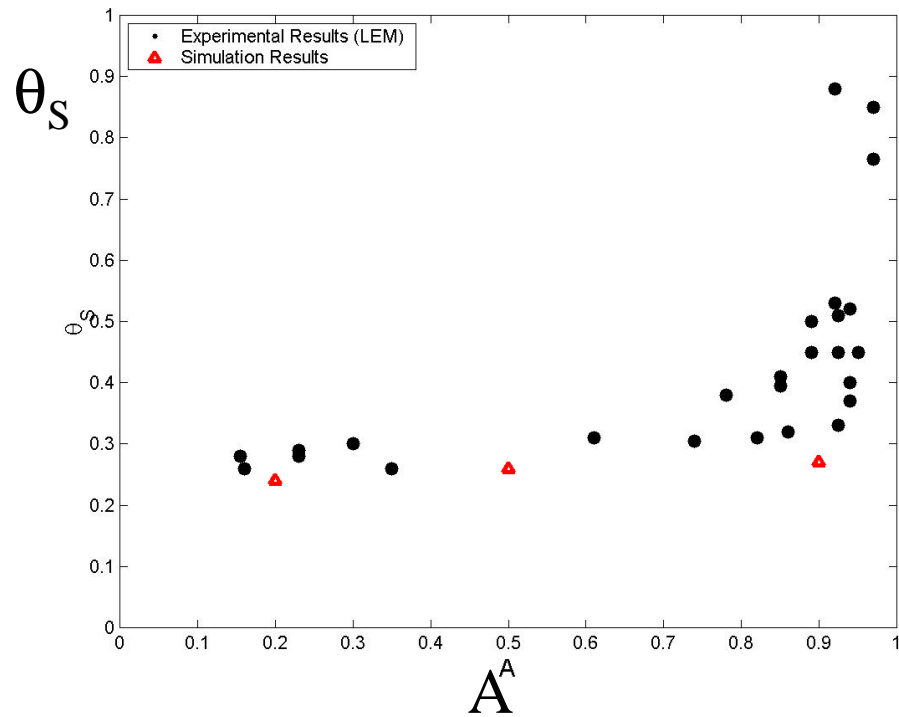
A

LEM Experimental Results Vs. 3D Simulations

Alpha - Spike

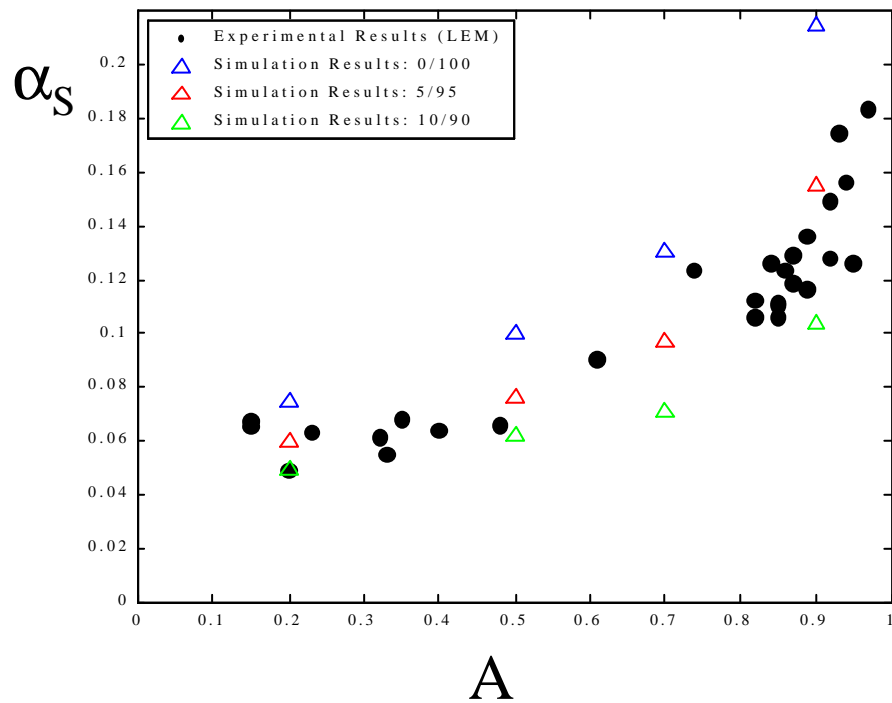


Theta - Spike

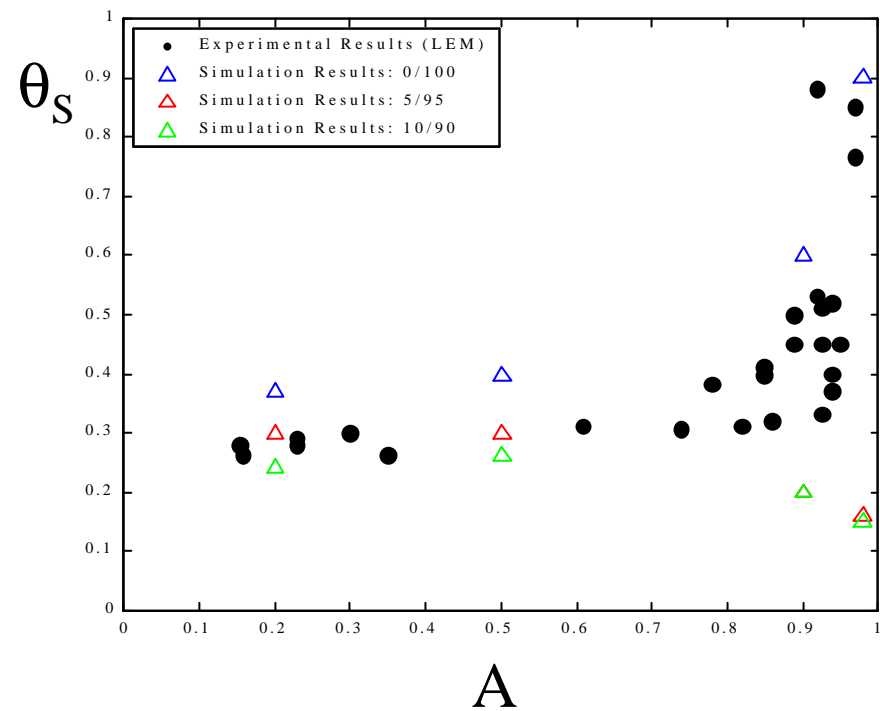


Sensitivity of q_s to percentage criterion

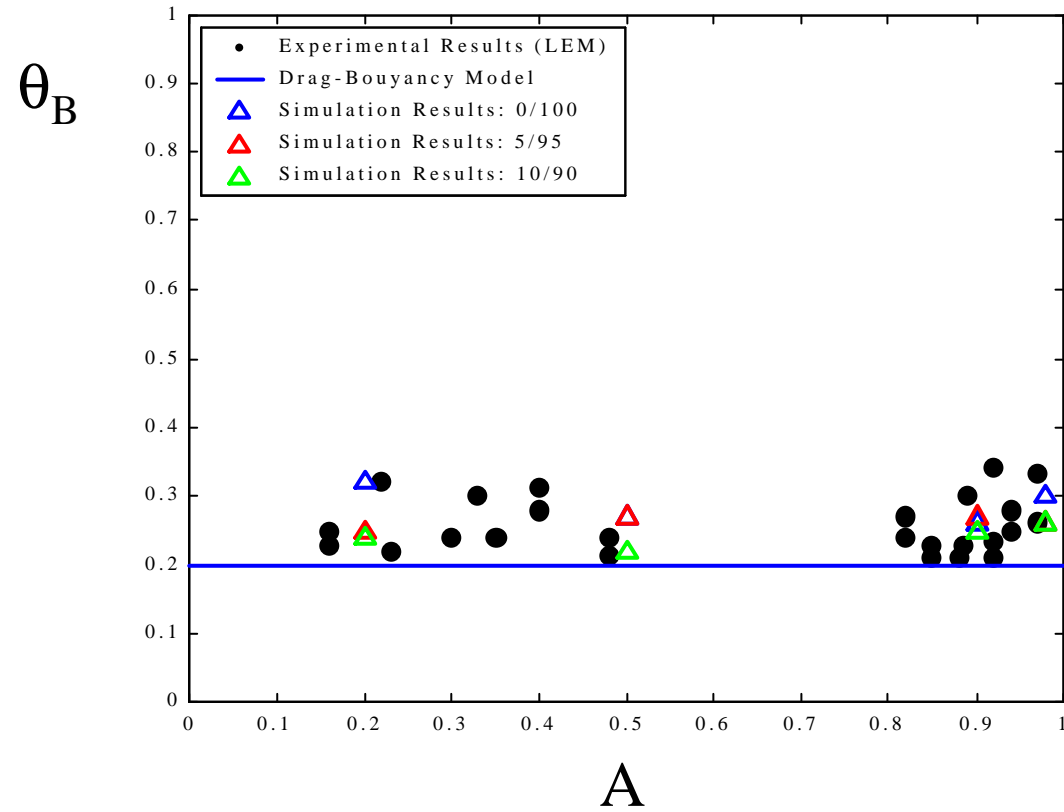
Alpha - Spike



Theta - Spike



q_B is not sensitive to percentage criterion



Outline

The Drag-Buoyancy Model for Bubbles

2D Statistical Model

Full 3D Numerical Simulations

3D Statistical Model

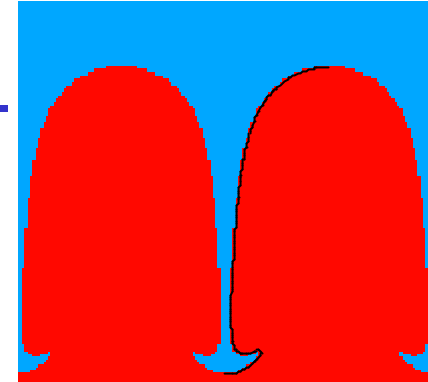
The Drag-Buoyancy Model for Spikes

Drag-Buoyancy Model for the Spike Front -

I. Single Mode

Since the spikes develop a rounded tip, one can apply the drag-buoyancy equation to them:

$$(\rho_h + C_a \rho_l) \dot{u}_s = (\rho_h - \rho_l) \cdot g - \frac{C_d}{\lambda} \rho_l \cdot u_s^2$$

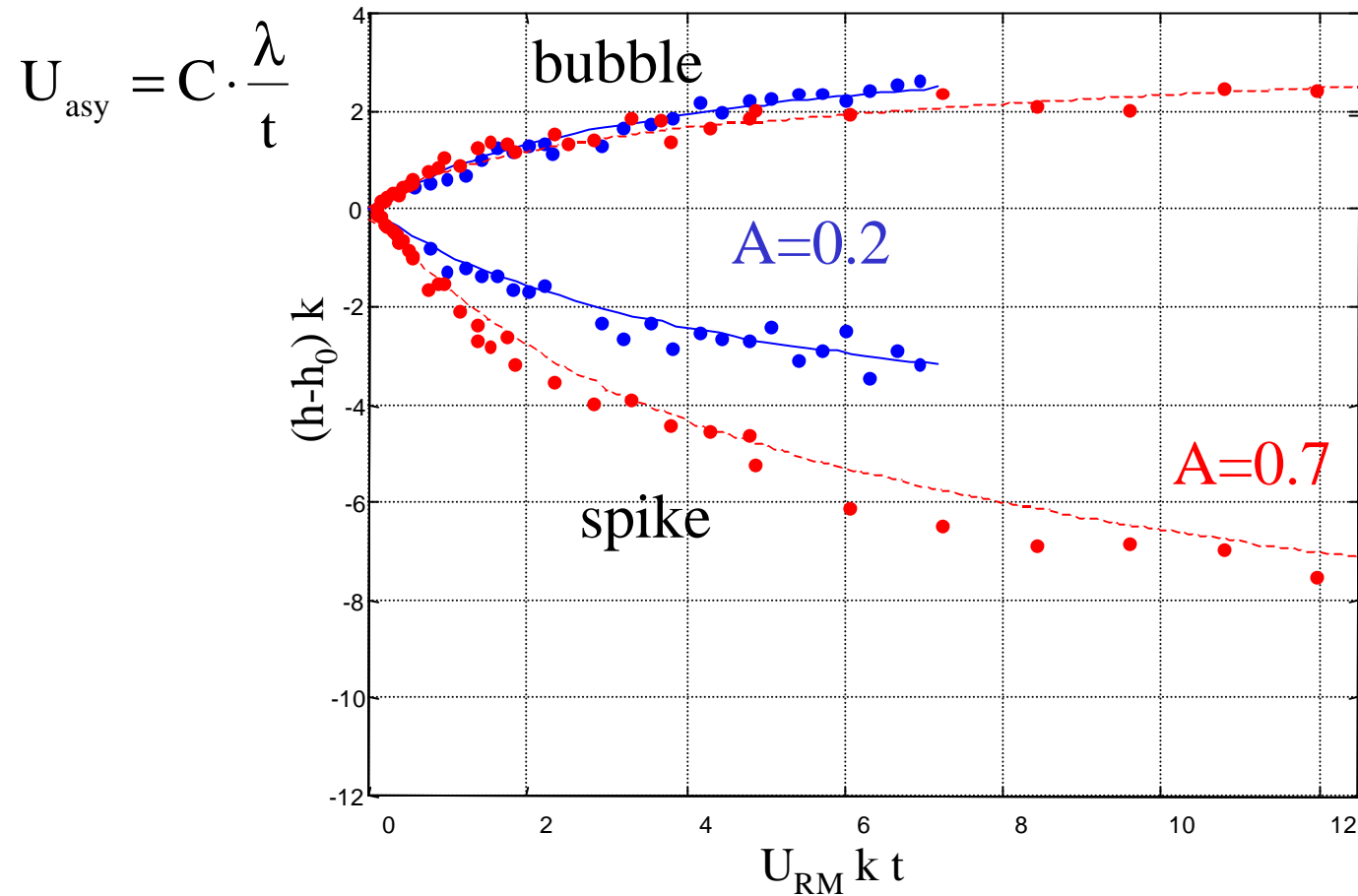


The spikes velocity is obtained using the assumptions:

$$C_{dS}(A) = C_{dB}(A) = \begin{cases} 6\pi & 2D \\ 2\pi & 3D \end{cases}$$

$$C_{aS}(A) = C_{aB}(A) = \begin{cases} 2 & 2D \\ 1 & 3D \end{cases}$$

Drag-Buoyancy Model for the Spike Front - I. Single Mode Shock Tube Experimental Results

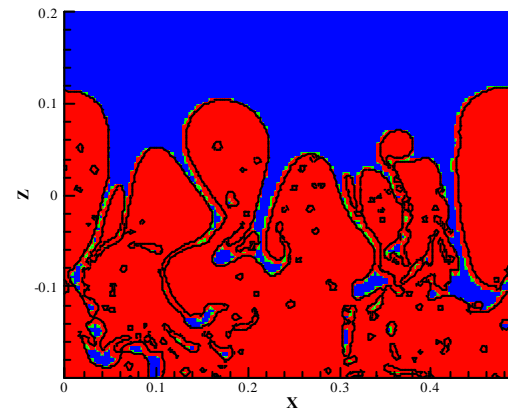
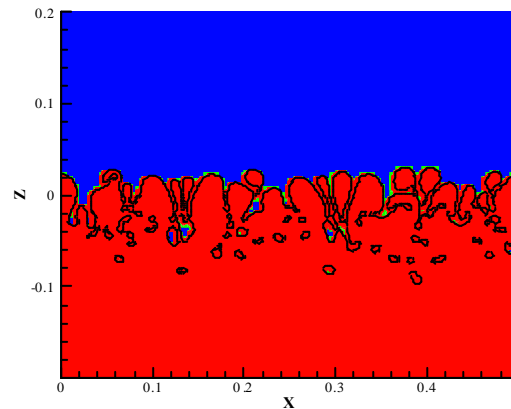


Drag-Buoyancy Model for the Spike Front - II. Multi Mode

1st assumption:

Periodicity of the spikes \equiv Periodicity of the bubbles

$$\lambda_S(t) \equiv \lambda_B(t)$$



2D RT Simulation with $A=0.9$

Drag-Buoyancy Model for the Spike Front - II. Multi Mode - RT

Na ve approach:

The ratio between the momentary velocities of the spikes and the bubbles equals the ratio between their asymptotic velocities:

$$\frac{u_s(t)}{u_B(t)} = \frac{2\alpha_s A g t}{2\alpha_B A g t} = \frac{\alpha_s}{\alpha_B} = \frac{u_s^{\text{asy}}}{u_B^{\text{asy}}} = \sqrt{\frac{1+A}{1-A}} = \sqrt{\frac{\rho_h}{\rho_l}}$$

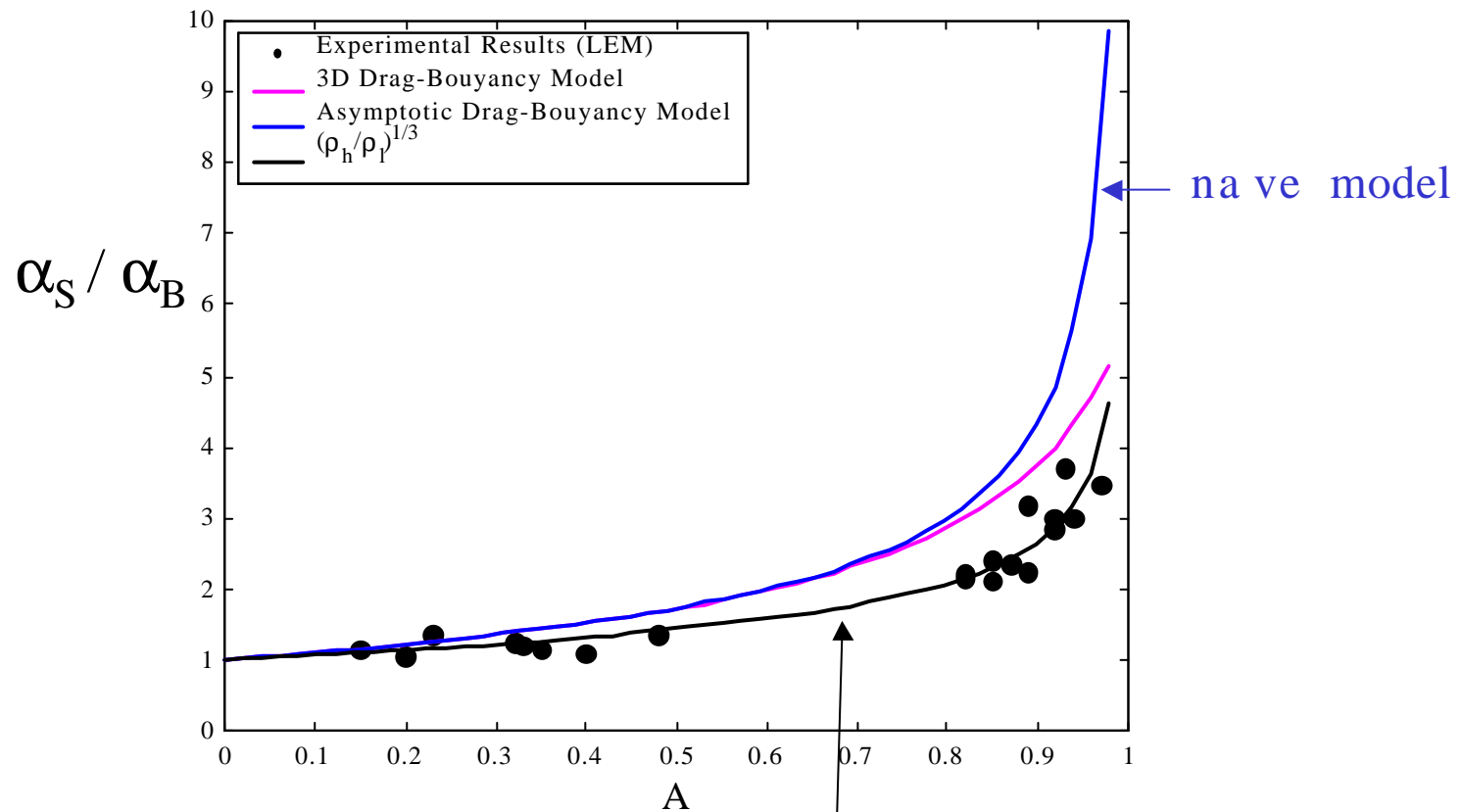
2nd assumption:

The ratio between the momentary velocities of the spikes and the bubbles equals the ratio between their velocities at the time t_b in which the bubbles height reaches its self-similar value: $h_B(t_b) = b\lambda$

$$\frac{u_s(t)}{u_B(t)} = \frac{2\alpha_s A g t}{2\alpha_B A g t} = \frac{\alpha_s}{\alpha_B} = \frac{u_s(t_b)}{u_B(t_b)} \cong \frac{u_s(t_b)}{u_B^{\text{asy}}}$$

Drag-Buoyancy Model for the Spike Front - II. Multi Mode - RT Results

3D:
$$\frac{\alpha_S}{\alpha_B} = \sqrt{\frac{1+A}{1-A}} \cdot \tanh \left[\sqrt{\frac{1-A}{1+A}} \cdot \cosh^{-1} \cdot \exp(\pi b(1+A)) \right]$$



$\alpha_S / \alpha_B \sim (\rho_h / \rho_l)^{1/3}$: Best fit by Dimonte et. Al.

Drag-Buoyancy Model for the Spike Front - II. Single Mode - RM

At late times, the bubble velocity in RM goes like λ/t .

At finite time t its velocity can be expressed by a time-dependent coefficient $\gamma(t)$:

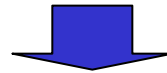
$$\frac{u_B(t)}{u_0} \propto \left(\frac{u_0 t}{\lambda} \right)^{\gamma_B(t)} \quad \gamma_B(t \rightarrow \infty) = -1 \quad \text{for all } A \text{ s}$$

$$\frac{u_S(t)}{u_0} \propto \left(\frac{u_0 t}{\lambda} \right)^{\gamma_S(t)} \quad \gamma_S(t \rightarrow \infty) = \begin{cases} -1 & A < 1 \\ 0 & A = 1 \end{cases}$$

Where, from assumption 1: $\langle \lambda(t) \rangle = \frac{h_B(t)}{b} \propto t^{\theta_B}$

Drag-Buoyancy Model for the Spike Front - II. Multi Mode - RM

$$\Rightarrow u_S^{\text{MM}} \propto t^{\gamma_S(t_b) \cdot (1 - \theta_B)} \quad \text{also:} \quad u_S^{\text{MM}}(t) = \frac{dh_S(t)}{dt} \propto t^{\theta_S - 1}$$



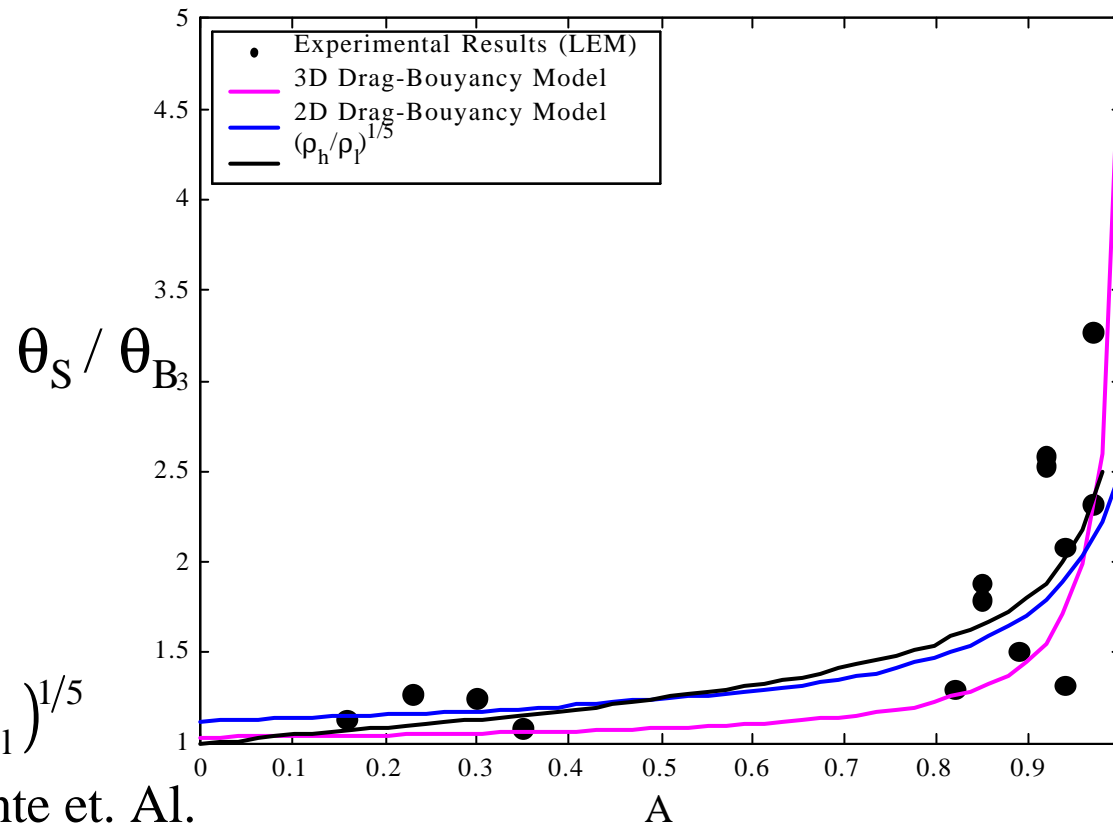
$$\boxed{\theta_S = 1 + \gamma_S(t_b) \cdot (1 - \theta_B)}$$

Na ve approach:

$$\gamma_S^{\text{asy}} = \begin{cases} 0 & A = 1 \rightarrow \theta_S = 1 \\ -1 & A < 1 \rightarrow \theta_S = \theta_B \end{cases}$$

Drag-Buoyancy Model for the Spike Front - II. Multi Mode - RM Results

3D:
$$\theta_S = 1 + \gamma_S(t_b) \cdot (1 - \theta_B) = 1 - \frac{\left(\frac{1-A}{1+A}\right) \cdot (\exp(\pi b(1+A)) - 1)}{1 + \left(\frac{1-A}{1+A}\right) \cdot (\exp(\pi b(1+A)) - 1)} \cdot (1 - \theta_B)$$



$$\alpha_S / \alpha_B \sim (\rho_h / \rho_l)^{1/5}$$

Best fit by Dimonte et. Al.

Summary

Good agreement between 3D drag-buoyancy model, statistical model, full 3D simulations and experimental results.

2D and 3D RT and RM scaling laws:

$$\text{RT: } h_{B(S)} = \alpha_{B(S)} A g t^2$$

$$\text{RM: } h_{B(S)} = a_{B(S)} t^{\theta_{B(S)}}$$

Geometrical effect results in different scaling parameters:

$$\alpha_B \approx 0.05 \quad \text{in 2D and 3D}$$

$$\theta_B \approx 0.2-0.3 \quad \text{in 3D}$$

$$\text{Vs. } 0.4 \quad \text{in 2D}$$

$$b \approx 0.75-1.5 \quad \text{in 3D}$$

$$\text{Vs. } 0.25-0.5 \quad \text{in 2D}$$

Spikes scaling laws are obtained from the drag-buoyancy model⁴¹.