Research in the Center for Radiative Shock Hydrodynamics (CRASH)





Center for Laser Experimental Astrophysics Research



R Paul Drake University of Michigan



Center for Radiative Shock Hydrodynamics

Department of Atmospheric Oceanic & Space Sciences

Applied Physics Program Department of Physics

Michigan Institute for Plasma Science and Engineering



Many individuals contribute to the CRASH Team

- Co-Principal Investigators
 - UM: James P. Holloway, Kenneth G. Powell, Quentin Stout
 - TAMU: Marvin L. Adams
- Participants
 - UM: Eight departments (Math, Stats + six in Engineering)
 - Ten instructional faculty
 - Eight research faculty
 - Twenty graduate students
 - Engineers, administrators, undergraduates
 - TAMU: Three departments (Nuclear, CompSci, Stats)
 - Six instructional faculty
 - Eight graduate students
 - Technical staff
 - Simon Frazer U.: Prof. Derek Bingham and one graduate student



We value our scientific and financial collaborators

Scientific collaborators (partial list):

LLE/Rochester – Knauer, Boehly, Nilson, Froula, Fiskel, others LLNL – Park, Remington, Glenzer, Fournier, Doeppner, Miles, Ryutov, Smalyuk, Hurricane, others LANL – Montgomery, Lanier, others Florida State – Plewa France – Bouquet, Koenig, Michaut, Loupias, others Britain -- Lebedev Texas – Wheeler Arizona – Arnett, Meakin Negev – Shvarts, Malamud Chicago – Abarzhi, others Financial collaborators: CRASH: Predictive Science Academic Alliance Program, DOE/NNSA/ ASC (grant DE-FC52-08NA28616)

CLEAR:

Joint HEDLP program (grant DE-FG52-04NA00064) National Laser User Facility (grant DE-FG03–00SF22021) DTRA grant HDTRA-1-10-0077 Los Alamos Nat. Lab. Laboratory for Laser Energetics Past support: Lawrence Livermore Nat. Lab. Naval Research Lab.



CRASH is focused on *predictive science*

- What CRASH is about:
 - Our goal is to test methods that evaluate our predictive capability to model complex behavior
 - The predictor is a multiphysics computer code
 - Radiation hydrodynamic experiments are modeled
 - Our approach is to predict the behavior of a more complex system based on measurements of simpler systems
- This talk:
 - Our radiative shock system and experiments
 - The CRASH code
 - Predictive science studies



Shocks become radiative when ...

 Radiative energy flux would exceed incoming material energy flux



where post-shock temperature is proportional to u_s^2 .

 Setting these fluxes equal gives a threshold velocity of 60 km/s atmospheric-pressure xenon:

			CRASH
Initial shock velocity	200 km/s	Typ. radiation temp.	50 eV
Density	6.5 mg/cc	Initial ion temperature	2 keV
Material	xenon gas		

Our simple system is a radiative shock in a circular tube

- 1 ns, 3.8 kJ laser irradiates Be disk
- Drives shock down Xe-filled tube
- Radiation ablates wall of tube -> wall shock
- **Ongoing CRASH** experiments chosen first to improve then to test predictive capability



We have used radiography to investigate the lateral structure of these shocks

- Bayesian analysis of tilt gives compression ~ 22
 - Doss HEDP, A&SS 2010
- Shock-shock interactions give local Mach number
 - Doss PoP 2009

- Shape of entrained flow reveals wave-wave dynamics
 - Doss PoP 2011
- Thin layer instability; scaling to supernova remnants
 - Doss thesis & to be pub.



We are also making other measurements

• Shock breakout from the Be disk



• X-ray Thomson scattering







- Papers in prep
 - Kuranz et al.
 - Stripling et al.
 - Visco et al.
 - Huntington et al.



We simulate the experiments using the CRASH code

- Dynamic adaptive AMR
- Level set interfaces
- Self-consistent EOS and opacities or other tables
- Multigroup-diffusion radiation transport
- Electron physics and fluxlimited electron heat conduction
- Laser package
- Ongoing
 - Multigroup preconditioner
 - I/O performance upgrade

3D Nozzle to Ellipse @ 13 ns



Material & AMR



Log Density



Log Electron Temperature



Log Ion Temperature

CRASH code: Van der Holst et al, Ap.J.S. 2011



The CRASH 3.0 simulation of the simple experiment reproduces many observed aspects



time = 0.0 ns

All physics, 10 hours on 100 cores



The shock at 13 ns looks much like the data











Our complex system drives such a shock into an elliptical tube



Our work in predictive science revolves around inputs and outputs of the code



X - Experiment parameters
θ - Physical Constants
N - Numerical Parameters

Y - Results passed forward and/or analyzed with data by statistical methods



Our inputs and outputs reflect the specifics of our experimental system

Inputs

- Experimental (x)
 - Laser energy
 - Be disk thickness
 - Xe fill gas pressure
- Model parameters (θ)
 - Vary with model
 - Examples:
 - electron flux limiter, laser energy scale factor,
 - opacity or group scale factor
- Form of model
 - e.g. 2D vs 3D

- Outputs (y)
 - Integrated Metrics
 - Shock location (SL)
 - Axial centroid of dense Xe (AC)
 - Area of dense Xe (A)
 - Radial moments
 - Shock breakout time (BOT)



1400 1600 1800 2000 2200 2400 Target Coord. X (μm)



We draw conclusions by comparing run sets in which we vary the inputs with experimental outputs

- Typical multi-D run sets are 128 runs, limited by available cycles
- Run sets are space-filling Latin Hypercube designs
- Current analysis is via Gaussian-process Bayesian modeling

(Sorry for the opaque jargon – no time to explain)



We use a model structure for calibration, validation & uncertainty assessment



Kennedy & O'Hagan 2000, 2001



Flux limiter is an uncertain model parameter

- Need to evaluate probability distribution of such parameters
- This can represent calibration or tuning
- If the residual discrepancy is small, we get calibration
- If not, we get tuning



We combine such models ...

- In sequence:
 - One set of experiments can be used to calibrate parameter probability distributions
 - These can be used in another model to predict
- Jointly:
 - Allows use of cheap and expensive models
 - Model-model discrepancy corrects the cheap model to the expensive one
 - Use a field-model discrepancy as before
 - Jointly fit both and calibrate/tune







We are now working to combine complex models and predict our complex experiment

- Combine predictions from multiple integral models that are not a strict hierarchy
 - Faster running models can help explore the dependence on the input variables
 - Jointly use multigroup 2D and Gray 3D
- Tune faster running models to slower, better models
 - e.g. 2D circular tube to 3D oval tube
- Better understand calibration in combined models
- Propose best next sets of runs to optimally reduce expected integrated MSE in fitting
- Predict year 4/5 experiment



Thanks



http://aoss-research.engin.umich.edu/crash/

