

# Challenges and progress in physical tomographic reconstruction of light doses for volumetric additive manufacturing

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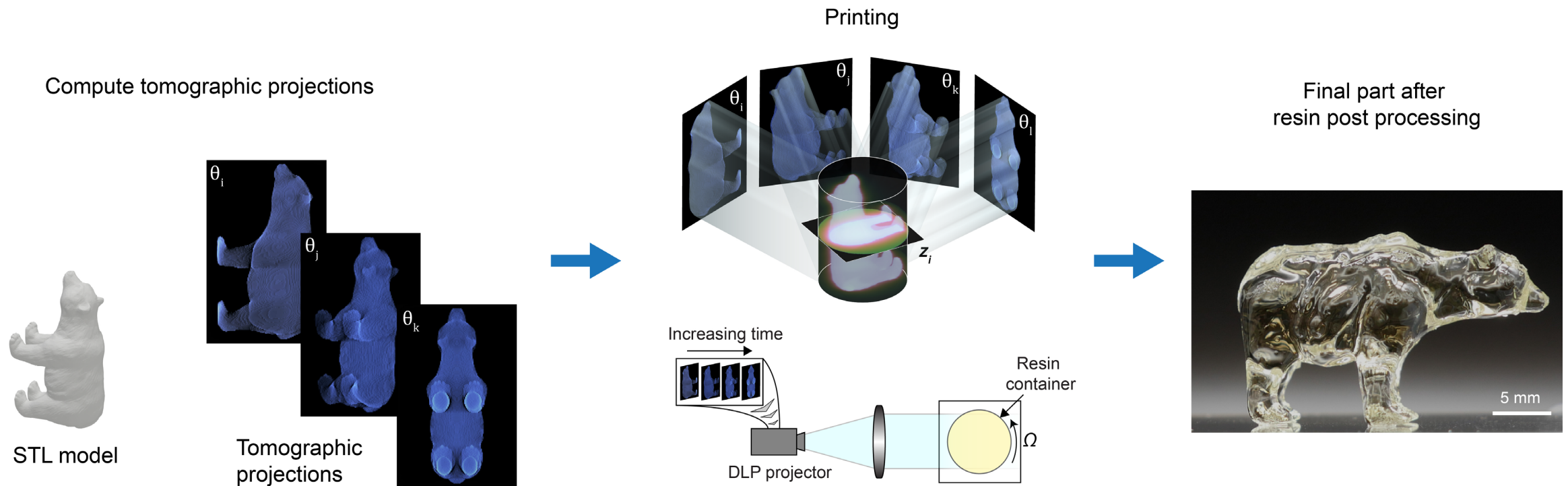
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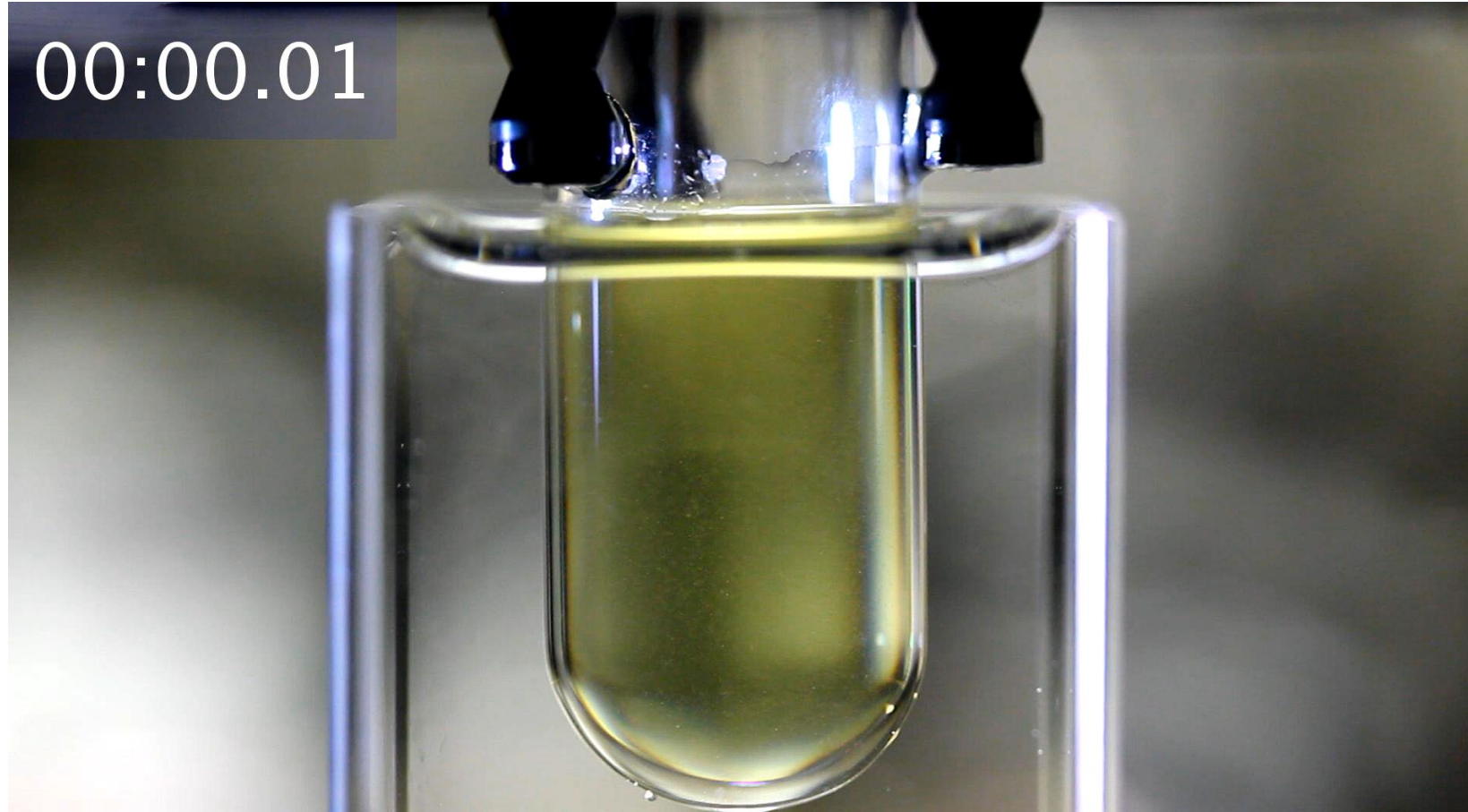
\*hkt@berkeley.edu

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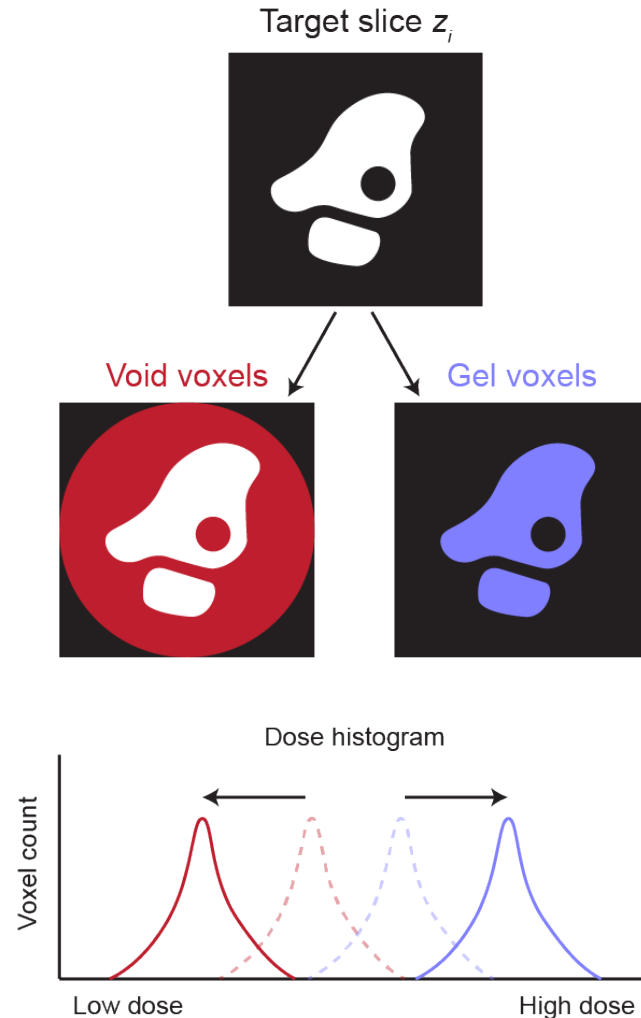
# Computed Axial Lithography – a tomographic volumetric additive manufacturing (VAM) technique



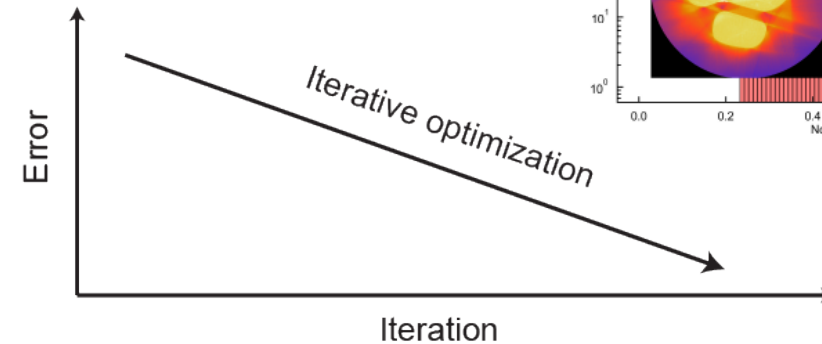
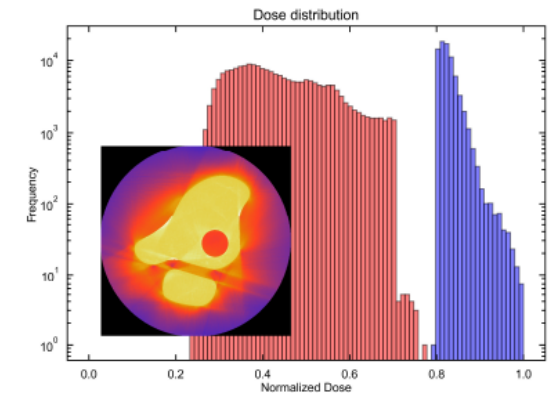
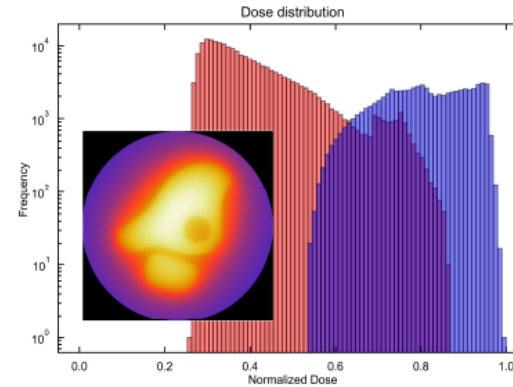
# Computed Axial Lithography – a tomographic volumetric additive manufacturing technique



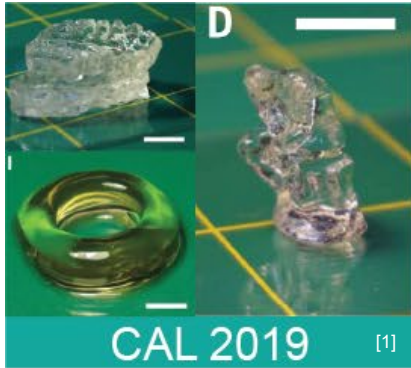
# Why do we need to optimize the digital light projections?



Goal of optimization is to push gel and void dose distributions apart to achieve **high dose contrast**



# Various optimization techniques have been developed since the introduction of tomographic VAM



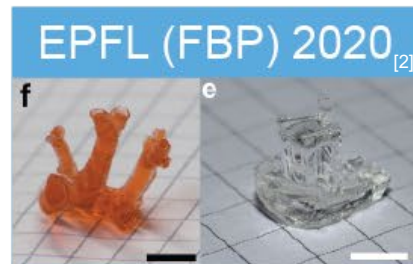
Filtered and positivity constrained backprojection, no optimization



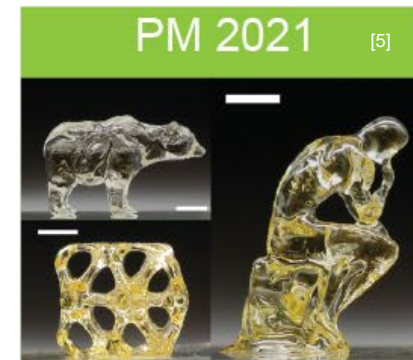
Analytic penalty (L1) minimization (PM) through L-BFGS quasi-Newton algorithm



Heuristic penalty (L1) minimization through finite difference approximate gradient descent



Analytic least squares (L2) minimization through conjugate gradient descent



Robust object-space model optimization (OSMO) through zeroth order method

[1] B. Kelly, I. Bhattacharya, H. Heidari et al. *Science*. **363**. (2019)  
 [2] D. Loterie, P. Delrot, C. Moser. *Nat. Comms*. **11**. (2020)  
 [3] C. Cook, E. Fong, J. Schwartz et al. *Adv. Mat*. **32**. (2020)  
 [4] M. Shusteff, K. Champley, E. Fong et al. US Pat. No. 20220011742  
 [5] I. Bhattacharya, J. Toombs, H. Taylor. *Additive Manufacturing*. **47**. (2021)  
 [6] C. Rackson, K. Champley, J. Toombs et al. *Additive Manufacturing*. **48**. (2021)  
 [7] J. Toombs, M. Luitz, C. Cook et al. arXiv: 2110.01651 (2021)

# Tomographic VAM algorithms span 0<sup>th</sup> to 2<sup>nd</sup> order methods with varying complexity

	CAL 2019	EPFL (FBP) 2020	LTT 2020	PM 2021	OSMO 2021
Optimization space	Projection	NA	Projection	Projection	Object
Analytical cost function	No	NA	Yes	Yes	NA
Order	1 <sup>st</sup>	NA	2 <sup>nd</sup>	2 <sup>nd</sup>	0 <sup>th</sup>
$p$ ( $\ell_p$ -norm)	1	NA	2	1	NA
Computational complexity (2D)*	$\mathcal{O}(n^2)$	$\mathcal{O}(n^2 \log n^2)^*$	Unknown	$\mathcal{O}(mn^2)^*$	$\mathcal{O}(n^2)$
Space complexity (2D)*	$\mathcal{O}(n^2)$	$\mathcal{O}(n^2)$	Unknown	$\mathcal{O}(mn^2)^*$	$\mathcal{O}(n^2)$
Open-source	Yes	Yes	No	Yes	Yes

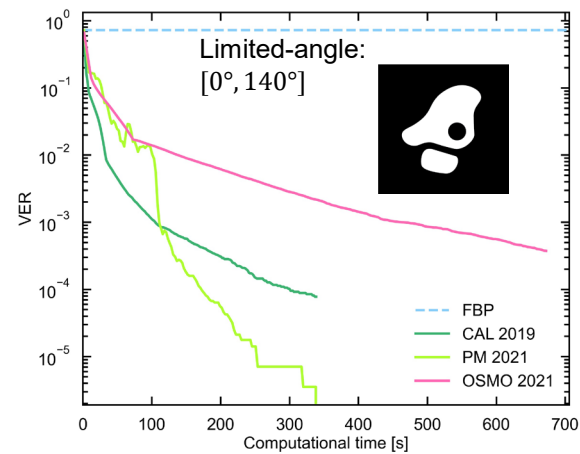
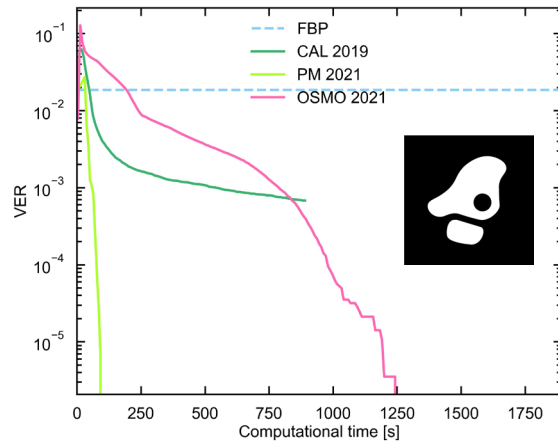
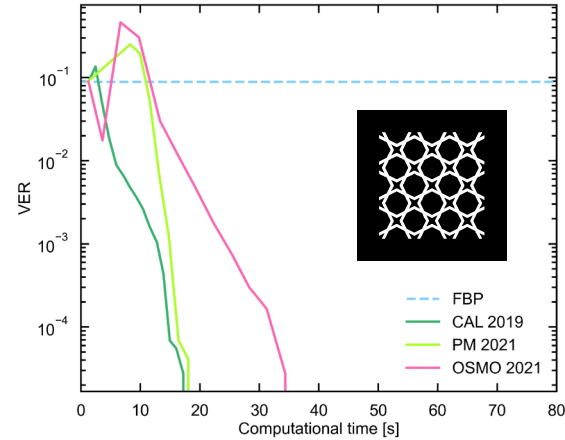
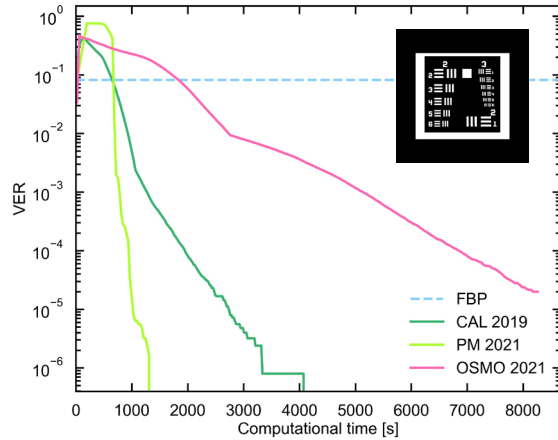
\*  $n$  is # of pixels/voxels on a side of the target matrix

\* Not an iterative method, only requires forward and backprojection and 1 FFT and 1 IFFT

\*  $m$  is # of historical steps stored to reconstruct approximate inverse Hessian in L-BFGS algorithm



# Performance trends

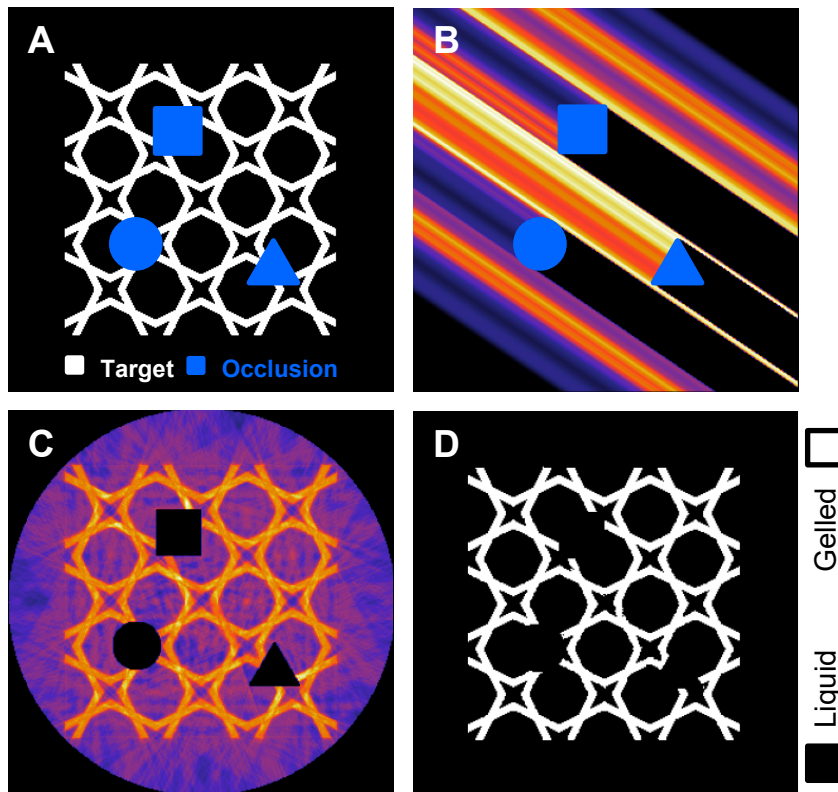


## Conclusions

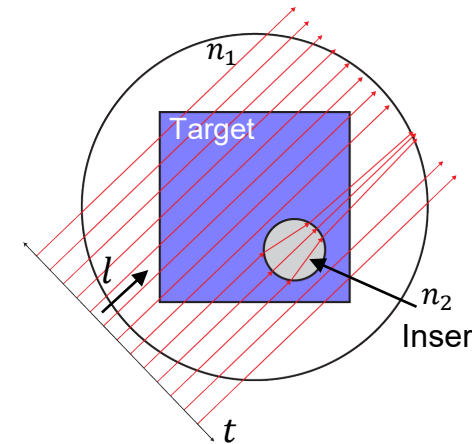
- All methods converge to solutions several orders of magnitude better than no optimization (FBP)
- First order methods (CAL 2019, OSMO 2021) are robust but require more iterations to converge
- High convergence rate of second order methods (PM 2021) may come at the cost of increased computational and space complexity for large 3D problems

# What about more challenging physical tomographic reconstruction? — A progress survey in multimaterial tomographic VAM — overprinting

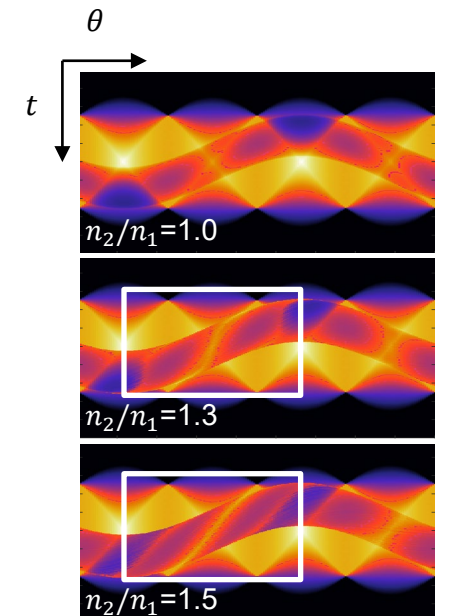
*Opaque insert/occlusion*



*Transparent (refractive) insert/occlusion*



$$[1] \quad \frac{d}{dl} \left( n \frac{dt}{dl} \right) - \frac{\partial n}{\partial t} = 0$$



Sinogram in region affected by the refractive insert is distorted

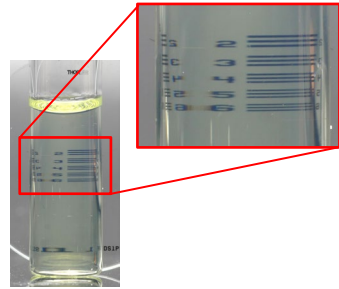
[1] C. Rackson, K. Champley, J. Toombs et al. *Additive Manufacturing*. **48**. (2021)

[2] Seron et al., *Comp. & Graphics*. **29**, (2005).

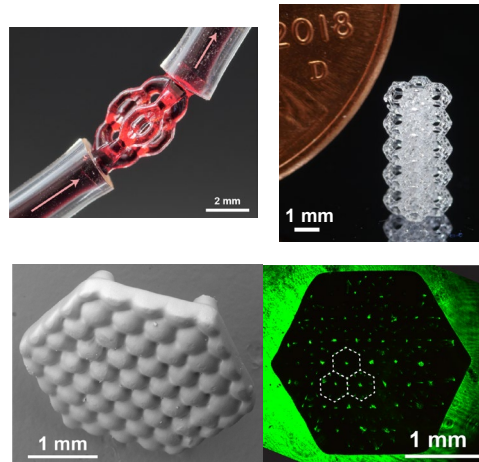
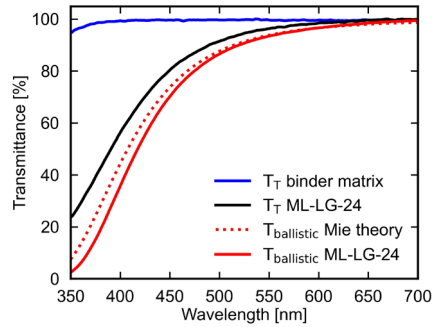


# A progress survey in multimaterial tomographic VAM — scattering

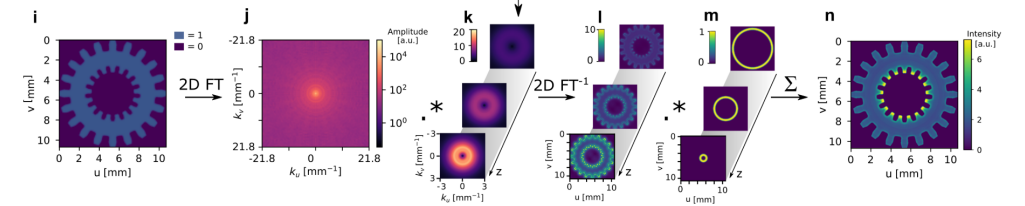
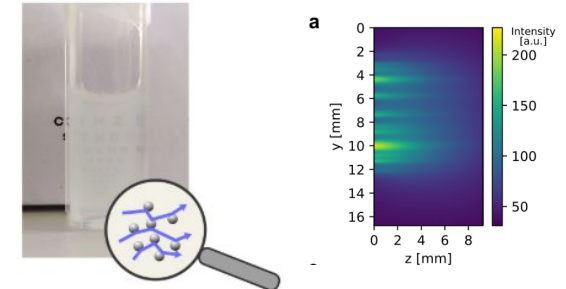
$$n_{particle} \approx n_{liquid}$$



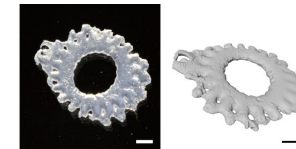
Refractive index matching  
 $n_{silica} \approx n_{precursor}$  leads to small scattering component of total transmission



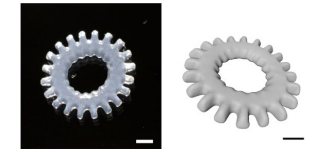
$$n_{particle} \neq n_{liquid}$$



Without precompensation



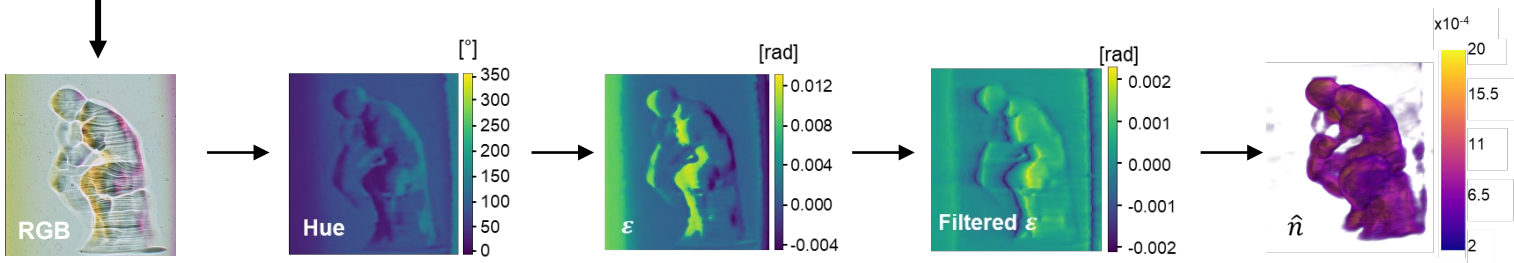
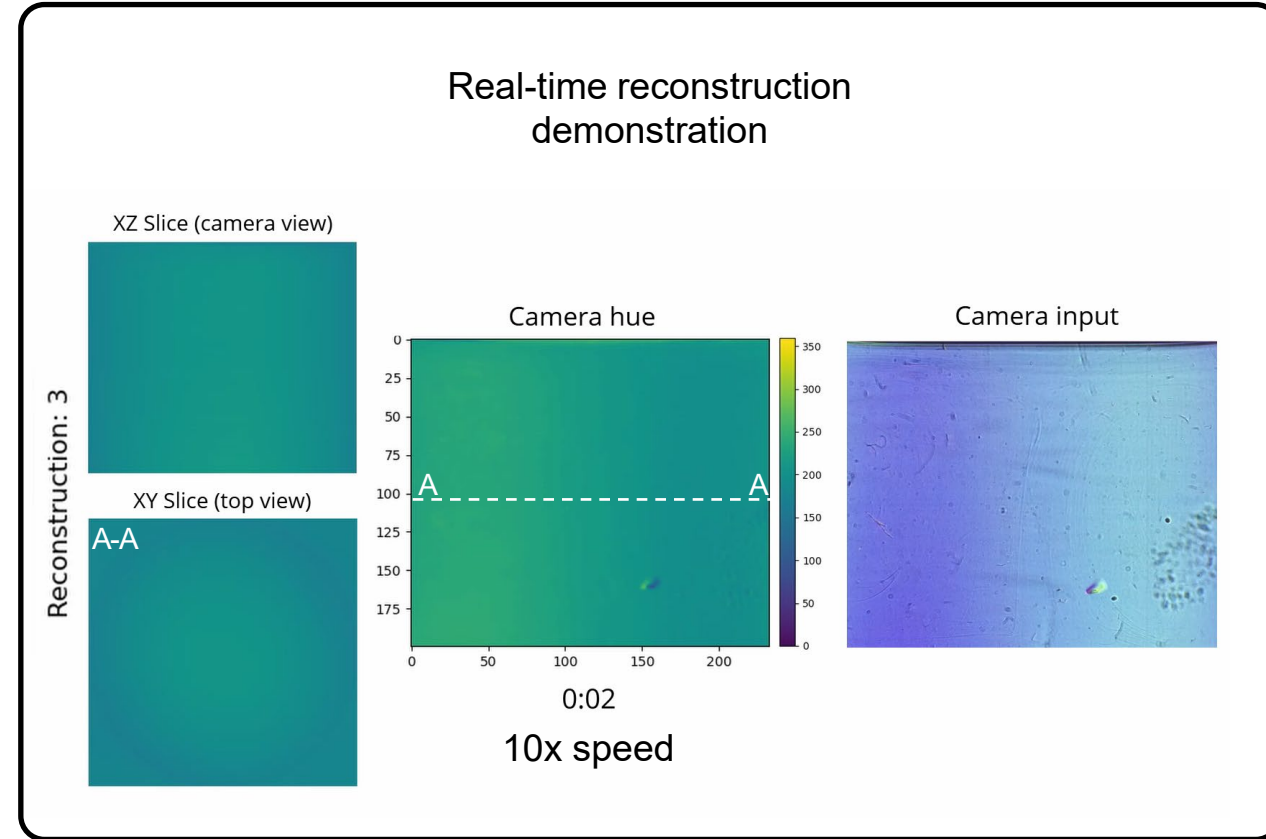
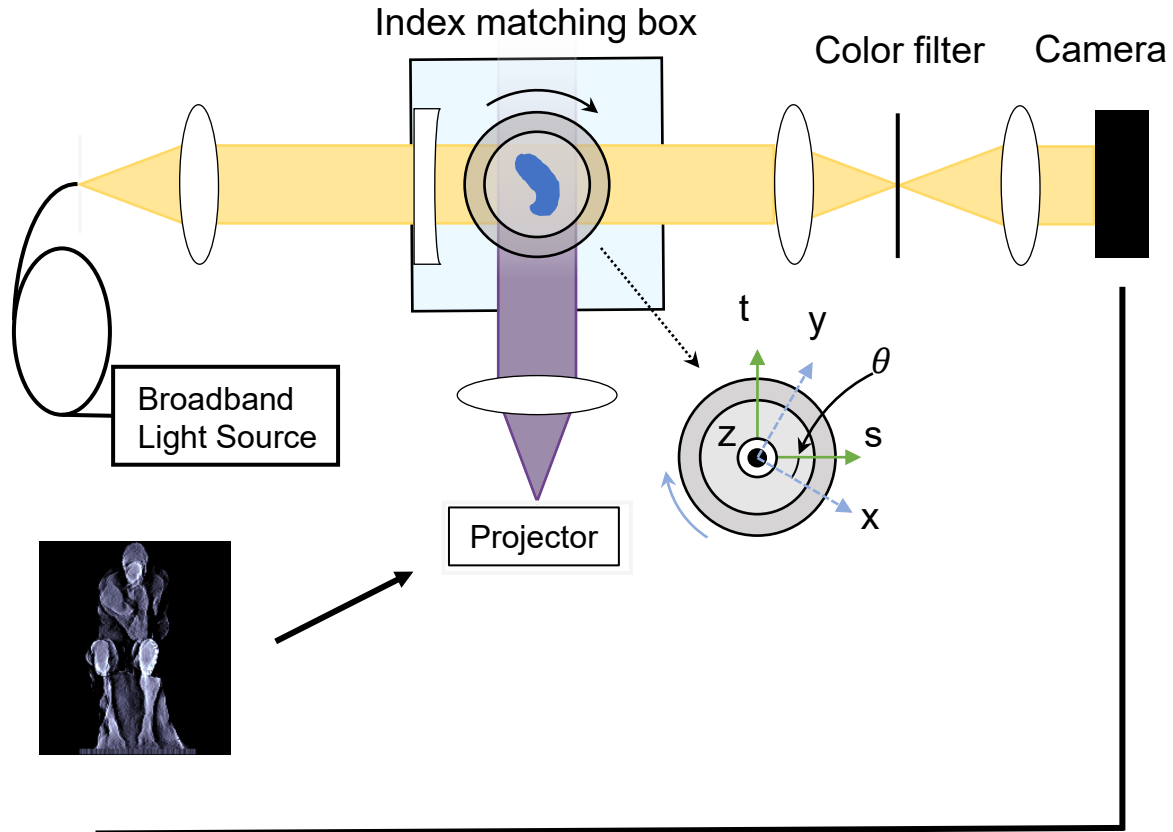
With precompensation



[1] J. Toombs, M. Luitz, C. Cook et al. arXiv: 2110.01651 (2021)

[2] J. Madrid-Wolff, A. Boniface, D. Loterie et al. arXiv: 2105.14952 (2021)

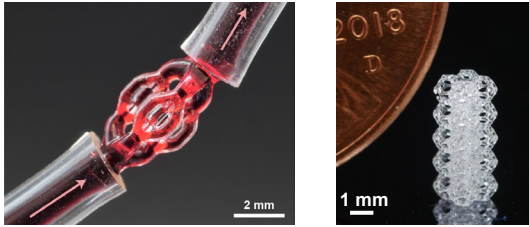
# When modeling the physical process becomes difficult Color Schlieren Tomography paves the way to real-time monitoring and process feedback



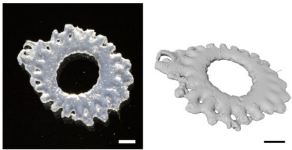
[1] C. Li, J. Toombs, H. Taylor (2021)

# Conclusions

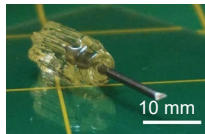
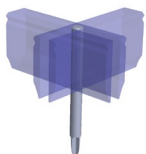
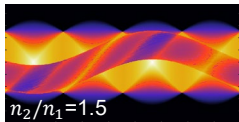
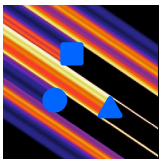
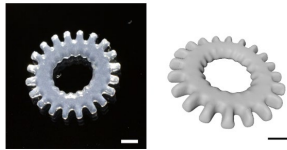
New forward/inverse light propagation models for multimaterial printing



Without precompensation

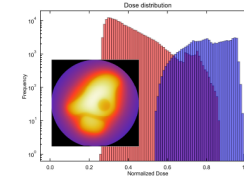
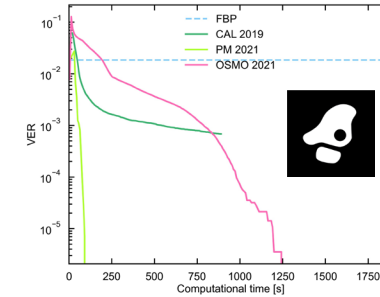


With precompensation

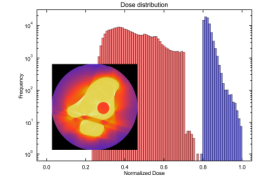


Tomographic  
VAM

Optimization for improved optical dose patterns



iterative  
optimization



In-situ metrology/computational imaging for process feedback and monitoring

