

**Fructose and the liver:
More than just extra calories?**

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STOPNASH Symposium, Washington, DC, Oct. 7, 2015

• **No disclosures**

Then

10:53:48:04

TS Media

2B105AD23_xxx

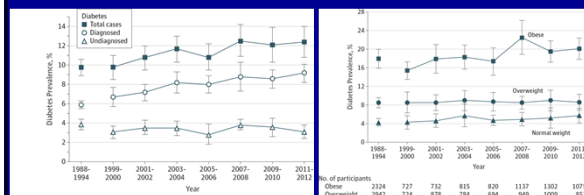
00:14:00:23

Dr. Fred Stare
Chair, Harvard School of Public Health
60 Minutes, 1970

Now

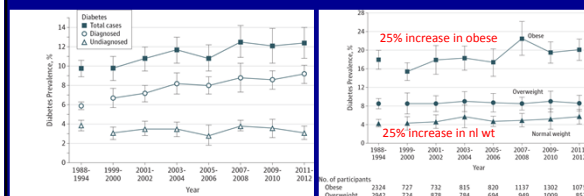
This video is from the WebMD Video Archive

Secular trend in diabetes among U.S. adults, 1988-2012

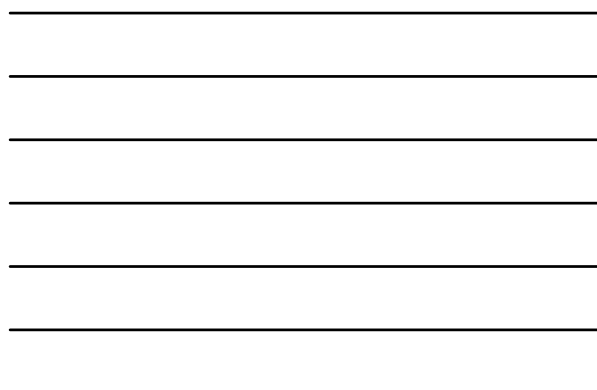


JAMA 314:1021, 2015, doi:10.1001/jama.2015.10029

Secular trend in diabetes among U.S. adults, 1988-2012



JAMA 314:1021, 2015, doi:10.1001/jama.2015.10029



Criticism from the nutritional community

Nutrition Research Reviews (2014), 27, 119–130 doi:10.1017/S0954579414000003
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Misconceptions about fructose-containing sugars and their role in the obesity epidemic

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University of Lausanne, Department of Physiology, 7 Rue du Bugnon, CH-1005, Lausanne, Switzerland

| | |
|------------------|---|
| Vincent Van Buul | Product Development Coordinator, Gruma/Mission Foods |
| Luc Tappy | Recipient of Numerous Fructose Grants, Nestle |
| Fred Brouns | Cargill Manager of Nutritional Sciences until 2008, now runs nutrition and health consulting firm |

Toxicity:

The degree to which a substance can damage an organism

- Does not distinguish acute vs. chronic toxicity

Requisites:

- Must be an "independent risk factor"
- Exclusive of calories
- Exclusive of obesity
- Must establish causation

Criticisms of Fructose Toxicity

- Animal models, not human studies
- Administration of excessive doses of fructose

Criticisms of Fructose Toxicity

- Animal models, not human studies
- Administration of excessive doses of fructose

WILL LIMIT DISCUSSION TO:
HUMAN DATA,
HUMAN CONSUMPTION,
AND IN DOSES ROUTINELY INGESTED

Sugar and Diabetes (Prospective Correlation)

SSB's and BMI-adjusted risk of diabetes in EPIC-Interact (Europe)

| Table 2 HRs (and 95% CIs) for type 2 diabetes according to type and amount of sweet beverage consumption in the EPIC-Interact study | | | | | |
|---|------------------------------------|---|---|--|-------------|
| Variable and model | <1 glass ^a /month HR | 1-4 glasses ^a /month HR ^b (95% CI) | >1-6 glasses ^a /week HR ^b (95% CI) | ≥1 glass ^a /day HR ^b (95% CI) | p for trend |
| Juices and nectars (median intake, g/day) | (0.0) | (17.3) | (100.0) | (338.3) | |
| No. cases | 5,837 | 1,702 | 3,625 | 720 | |
| Crude model | 1.00 (ref) | 0.88 (0.80, 0.98) | 0.89 (0.83, 0.94) | 0.97 (0.85, 1.11) | 0.64 |
| Adjusted model | 1.00 (ref) | 0.91 (0.80, 1.02) | 0.96 (0.88, 1.04) | 1.00 (0.87, 1.15) | 0.63 |
| Adjusted model+EI | 1.00 (ref) | 0.91 (0.81, 1.02) | 0.96 (0.88, 1.04) | 0.99 (0.86, 1.14) | 0.84 |
| Adjusted model+EI+BMI | 1.00 (ref) | 0.97 (0.86, 1.10) | 1.04 (0.96, 1.13) | 1.06 (0.90, 1.25) | 0.21 |
| Total soft drinks ^a (median intake, g/day) | (0.0) | (20.0) | (95.3) | (413.1) | |
| No. cases | 5,794 | 1,604 | 2,987 | 1,299 | |
| Crude model | 1.00 (ref) | 1.21 (1.07, 1.36) | 1.30 (1.18, 1.43) | 1.78 (1.55, 2.04) | <0.0001 |
| Adjusted model | 1.00 (ref) | 1.21 (1.07, 1.37) | 1.26 (1.13, 1.42) | 1.58 (1.35, 1.84) | <0.0001 |
| Adjusted model+EI | 1.00 (ref) | 1.21 (1.07, 1.37) | 1.27 (1.12, 1.43) | 1.59 (1.35, 1.88) | <0.0001 |
| Adjusted model+EI+BMI | 1.00 (ref) | 1.17 (0.97, 1.42) | 1.11 (0.98, 1.26) | 1.21 (1.05, 1.41) | 0.0005 |
| Sugar-sweetened soft drinks ^a (median intake, g/day) | (0.0) | (19.3) | (94.3) | (425.7) | |
| No. cases | 3,948 | 964 | 1,299 | 605 | |
| Crude model | 1.00 (ref) | 1.14 (0.97, 1.35) | 1.16 (1.05, 1.28) | 1.68 (1.40, 2.02) | <0.0001 |
| Adjusted model | 1.00 (ref) | 1.13 (0.97, 1.31) | 1.04 (0.94, 1.15) | 1.39 (1.16, 1.67) | <0.0001 |
| Adjusted model+EI | 1.00 (ref) | 1.12 (0.96, 1.31) | 1.04 (0.94, 1.15) | 1.39 (1.15, 1.68) | 0.001 |
| Adjusted model+EI+BMI | 1.00 (ref) | 1.19 (0.91, 1.56) | 1.07 (0.94, 1.21) | 1.29 (1.02, 1.63) | 0.013 |
| Artificially sweetened soft drinks ^a (median intake, g/day) | (0.0) | (18.3) | (99.0) | (260.0) | |
| No. cases | 5,242 | 609 | 894 | 291 | |
| Crude model | 1.00 (ref) | 1.09 (0.97, 1.23) | 1.52 (1.36, 1.69) | 1.84 (1.52, 2.23) | <0.0001 |
| Adjusted model | 1.00 (ref) | 1.10 (0.93, 1.29) | 1.46 (1.29, 1.65) | 1.93 (1.67, 2.24) | <0.0001 |
| Adjusted model+EI | 1.00 (ref) | 1.08 (0.93, 1.26) | 1.46 (1.29, 1.65) | 1.88 (1.64, 2.15) | <0.0001 |
| Adjusted model+EI+BMI | 1.00 (ref) | 1.05 (0.81, 1.35) | 1.38 (1.03, 1.85) | 1.13 (0.85, 1.52) | 0.24 |

Romaguera Bosch, et al. Diabetologia 56:1020, 2013

Associations between consumption of sugar sweetened beverages and fruit juice and incident type 2 diabetes: meta-analysis of prospective cohort studies

| Sugar Sweetened Beverages (n=17) | Not Adjusted for Adiposity: Relative Risk | I ² | Adjusted for Adiposity: Relative Risk | I ² |
|-----------------------------------|---|----------------|---------------------------------------|----------------|
| Meta-analysis, crude: | 1.25 (1.14 to 1.37) | 89 | — | — |
| +multivariable adjusted | 1.16 (1.09 to 1.28) | 89 | 1.13 (1.06 to 1.21) | 79 |
| +calibration for information bias | 1.43 (1.20 to 1.70) | 86 | 1.28 (1.12 to 1.46) | 73 |
| +calibration for publication bias | 1.42 (1.19 to 1.69) | 85 | 1.27 (1.10 to 1.46) | 73 |
| Fruit Juices (n=13) | Not Adjusted for Adiposity: Relative Risk | I ² | Adjusted for Adiposity: Relative Risk | I ² |
| Meta-analysis, crude: | 0.97 (0.90 to 1.06) | 79 | — | — |
| +multivariable adjusted | 1.05 (0.99 to 1.11) | 58 | 1.07 (1.01 to 1.14) | 51 |
| +calibration for information bias | 1.06 (0.98 to 1.14) | 49 | 1.10 (1.01 to 1.20) | 29 |
| +calibration for publication bias | Not detected | — | Not detected | — |

Imamura et al. BMJ. dx.doi.org/10.1136/bmj.h3576 (epub 21 July 2015)

Sugar and Diabetes (Econometric Analysis)

An international econometric analysis of diet and diabetes

Food and Agriculture Organization (FAO): FAOSTAT

Food Supply data in kcal/capita/day calculation:

Food Supply = \sum Supply Elements - \sum Utilization Elements =
(Production + Import Quantity + Stock Variation - Export Quantity)
- (Feed + Seed + Processing + Waste).
Only industrial waste factored in.

Extracted Food Supply data for 2000 and 2007:

| | |
|--------------------------------|--|
| Total Calories | Roots & Tubers, Pulses, Nuts, Vegetables |
| Fruits-Excluding Wine | Meat |
| Oils | Cereals |
| Sugar, Sugarcrops & Sweeteners | |

International Diabetes Federation (IDF)

2000 (1st ed) and 2010 (3rd ed)

The World Bank World Development Indicators Database

GDP expressed in purchasing power parity in 2005 US dollars for comparability among countries

Basu et al. PLoS One. e57873. 2013

An international econometric analysis of diet and diabetes

Total 175 countries; complete data for 154 countries (21 not different)

Basu et al. PLoS One. e57873. 2013

An international econometric analysis of diet and diabetes

Total 175 countries; complete data for 154 countries (21 not different)

Data monitoring and quality

- Generalized estimating equations
- Conservative fixed effects approach (Hausman test)
- Hazard model to control for selection bias (Heckman selection model)
- Longitudinal data to determine what preceded diabetes (Granger causality)
- Period effects controlled for secular trends that may have occurred as a result of changes diabetes detection capacity or importation policies.

Basu et al. PLoS One. e57873. 2013

An international econometric analysis of diet and diabetes

Total 175 countries; complete data for 154 countries (21 not different)

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- Period effects controlled for secular trends that may have occurred as a result of changes diabetes detection capacity or importation policies.

Controlled for:

| | |
|----------------|------------------------------------|
| GDP per capita | % population living in urban areas |
| Obesity | % of population over age 65 |
| | physical inactivity |

Basu et al. PLoS One. e57873. 2013

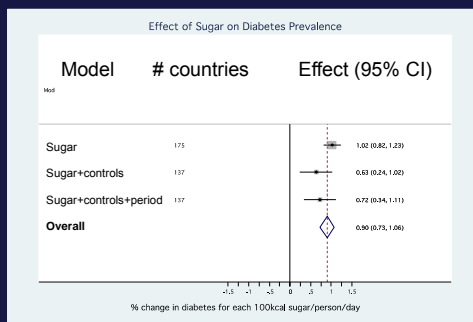
An international econometric analysis of diet and diabetes

Diabetes prevalence rose from 5.5% to 7.0% for 175 countries 2000-2010

Basu et al. PLoS One. e57873. 2013

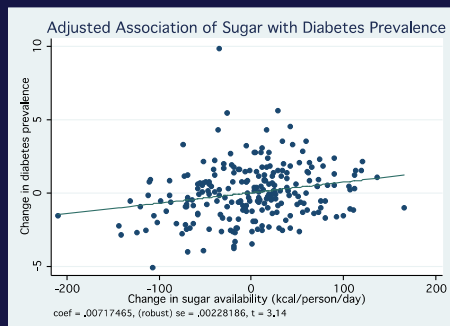
An international econometric analysis of diet and diabetes

Diabetes prevalence rose from 5.5% to 7.0% for 175 countries 2000-2010



Basu et al. PLoS One. e57873. 2013

An international econometric analysis of diet and diabetes



Basu et al. PLoS One. e57873. 2013

An international econometric analysis of diet and diabetes

Only changes in sugar availability predicted changes in diabetes prevalence

Every extra 150 calories increased diabetes prevalence by 0.1%

But if those 150 calories were a can of soda, diabetes prevalence increased 11-fold, by 1.1% (95% CI 0.03 — 1.71%, $p < 0.001$)

This study meets the Bradford Hill criteria for Causal Medical Inference:

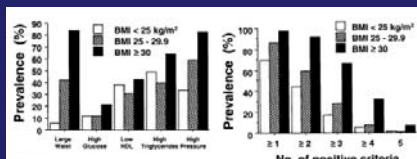
—dose —duration —directionality —precedence

We estimate that 25% of diabetes worldwide is explained by sugar

Basu et al. PLoS One. e57873. 2013

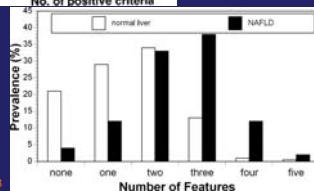
Sugar and Fatty Liver Disease

NAFLD and Metabolic Syndrome are congruent (if not the same)



Adults:
Marchesini et al. Hepatology 37:917, 2003

Children:
Schwimmer et al. Circulation 118:277, 2008



NAFLD is a primary predictor of T2DM in Korean adults

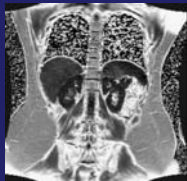
TABLE 2. OR for T2DM at 5-yr follow-up

| | T2DM -no./total no. (%) | | OR (95% confidence interval) | | |
|------------|-------------------------|----------------|------------------------------|-----------------------|--|
| | No fatty liver | Fatty liver | Unadjusted | Adjusted ^a | Adjusted ^a - baseline glucose |
| All | 54/8120 (0.7%) | 120/2971 (4%) | 6.29 (4.55-8.69) | 3.24 (2.19-4.78) | 2.05 (1.35-3.12) |
| Insulin | | | | | |
| Quartile 1 | 13/2468 (0.5%) | 8/307 (2.6%) | 5.05 (2.08-12.29) | 1.47 (1.23-9.79) | 1.98 (0.63-6.13) |
| Quartile 2 | 16/2262 (0.7%) | 6/511 (1.2%) | 1.67 (0.65-4.28) | 1.34 (0.46-3.87) | 0.71 (0.22-2.20) |
| Quartile 3 | 11/2002 (0.6%) | 22/768 (2.9%) | 5.34 (2.58-11.06) | 3.74 (1.59-8.84) | 2.92 (1.12-7.62) |
| Quartile 4 | 14/1388 (1.0%) | 84/1385 (6.1%) | 6.34 (3.58-11.21) | 5.31 (1.76-6.20) | 2.42 (1.23-4.75) |

^a Adjusted for age, gender, BMI, alcohol (grams per day), education (< 15 yr, • 15 yr), smoking (never or past, current), and exercise (< 1 time/week, • 1 time/week).

Sung and Kim, J Clin Endocrinol Metab 96:1093, 2011

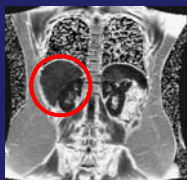
MRI Fat Fraction Maps



Obese

Low Liver Fat = 2.6%

MRI Fat Fraction Maps



Obese

Low Liver Fat = 2.6%

MRI Fat Fraction Maps

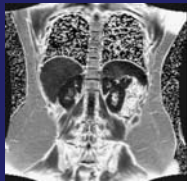


Obese
Low Liver Fat = 2.6%



Obese
High Liver Fat = 24%

MRI Fat Fraction Maps

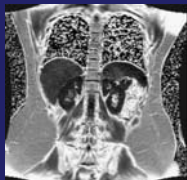


Obese
Low Liver Fat = 2.6%

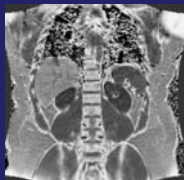


Obese
High Liver Fat = 24%

MRI Fat Fraction Maps



Obese
Low Liver Fat = 2.6%



Obese
High Liver Fat = 24%

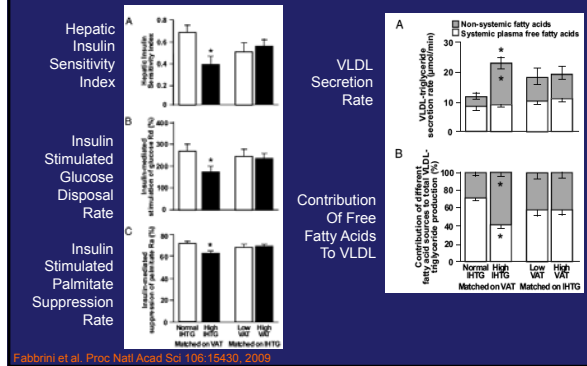


Thin
High Liver Fat = 23%

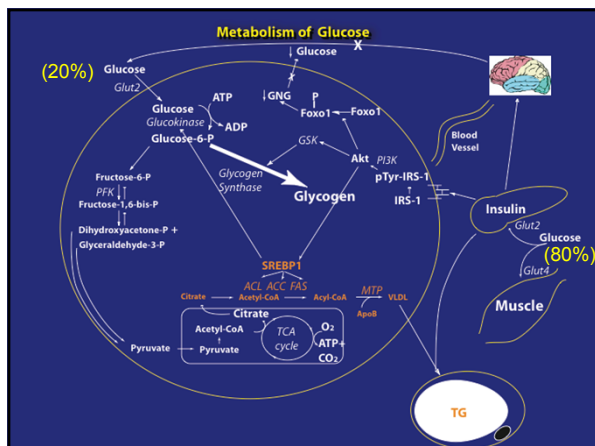
MRI Fat Fraction Maps



Intrahepatic fat explains metabolic perturbation better than visceral fat



Fabbrini et al. Proc Natl Acad Sci 106:15430, 2009



Can you name an energy source that is:

Can you name an energy source that is:

Not necessary for life

Can you name an energy source that is:

Not necessary for life

There is no biochemical reaction in the body that requires it

Can you name an energy source that is:

Not necessary for life

There is no biochemical reaction in the body that requires it

Is not nutrition

Can you name an energy source that is:

Not necessary for life

There is no biochemical reaction in the body that requires it

Is not nutrition

When consumed in excess it is toxic

Can you name an energy source that is:

Not necessary for life

There is no biochemical reaction in the body that requires it

Is not nutrition

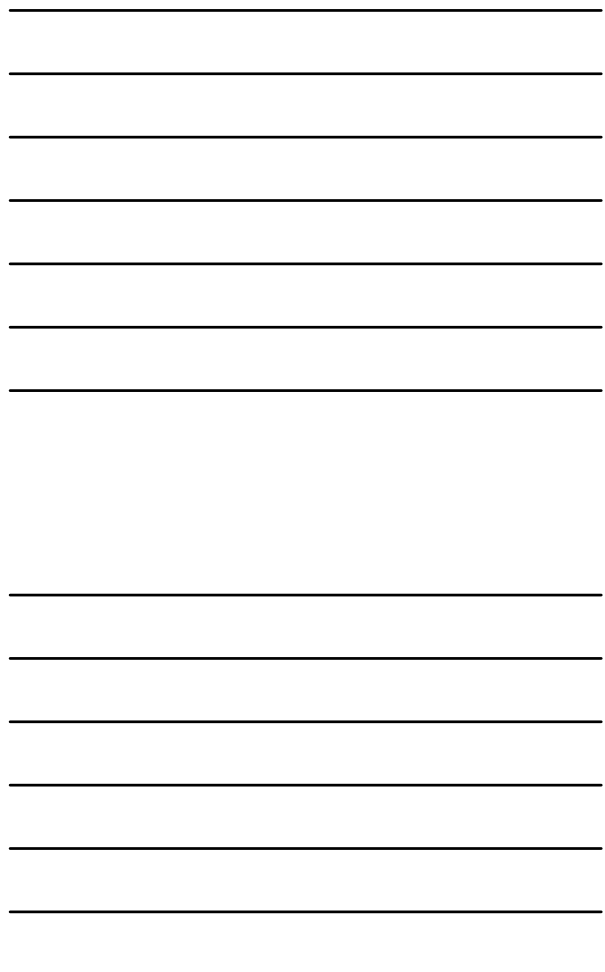
When consumed in excess it is toxic

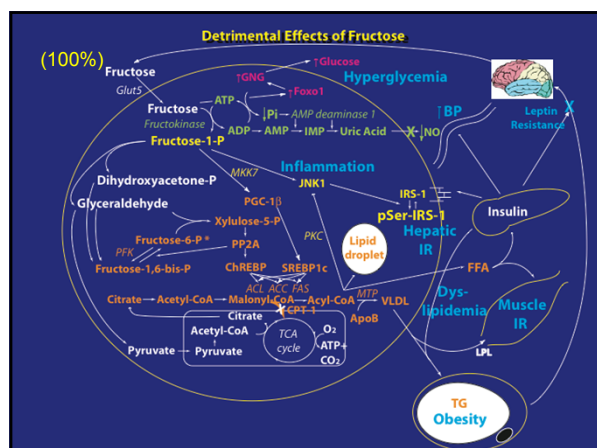
We love anyway

- Not necessary for life
- There is no biochemical reaction in the body that requires it
- Is not nutrition
- When consumed in excess it is toxic
- We love anyway

[illegible]

Alcohol?
Sugar?





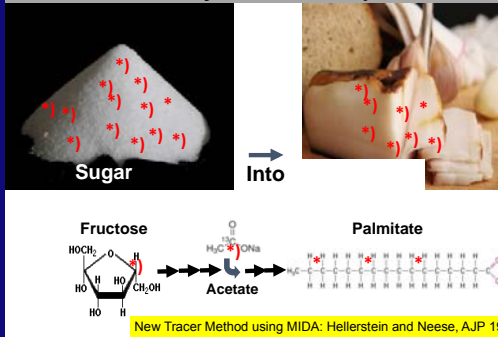
Sugar and Diabetes (Causation)

| | |
|--|--|
| <p>Current Press Releases</p> <ul style="list-style-type: none"> Press Release Archives ENDO Annual Meeting Media Webinars Journalism Award Science Writers Conference Endocrinology Glossary |  <h2 style="margin-top: 0;">Current Press Releases</h2> <div style="margin-top: 20px;"> <h3 style="font-size: 1.2em; margin: 0;">Restricting Fructose in Obese Latino and African American Children May Reduce Fat Accumulation in Their Liver</h3> <p style="margin: 10px 0 0 0;">March 05, 2015</p> <div style="display: flex; justify-content: space-between;"> <div style="width: 45%;"> <p>Contact: Aaron Lohr Director, Media Relations Phone: 202.971.3654 aalohr@endocrine.org</p> </div> <div style="width: 45%;"> <p>Contact: Jenni Glenn Gingery Associate Director, Media Relations Phone: 202.971.9655 jgingery@endocrine.org</p> </div> </div> <p style="margin-top: 20px;">San Diego, CA. – In obese Latino and African American children, restricting dietary fructose, but not</p> </div> |
|--|--|

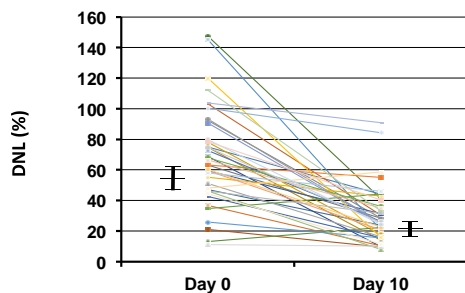
Strategy

- Isocaloric fructose restriction x 9 days in children who are habitual sugar consumers
- No change in weight
- Substitute complex carbs for sugar
- Maintain baseline macronutrient composition of the diet
- Study in PCRC at Day 0 and Day 10
- Assess changes in organ fat, *de novo* lipogenesis, and metabolic health

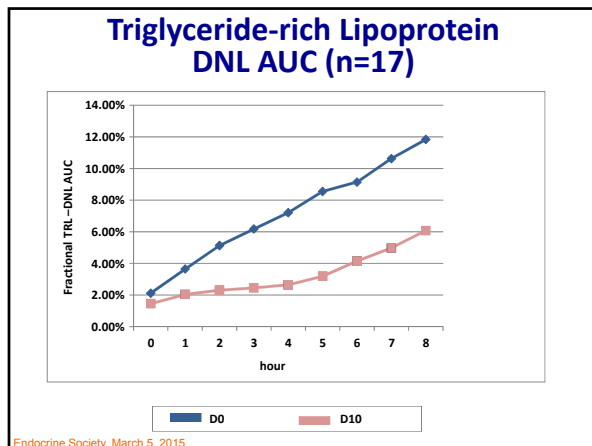
DNL is the Conversion of Dietary Carbohydrates into Lipids

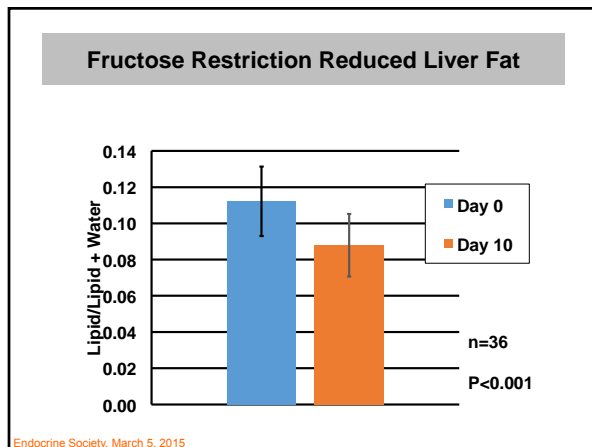


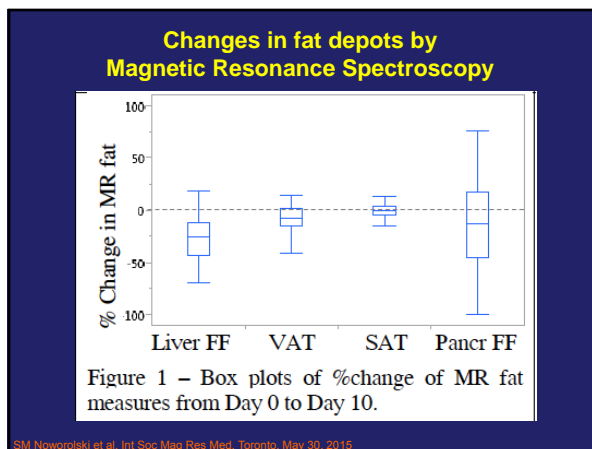
DNL AUC Pre and Post Fructose Restriction



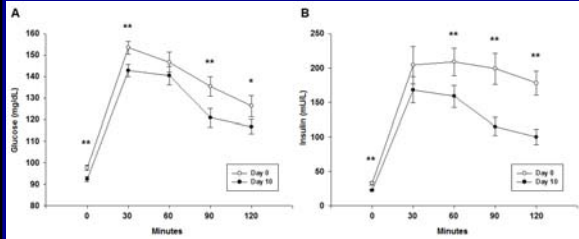
Endocrine Society, March 5, 2015







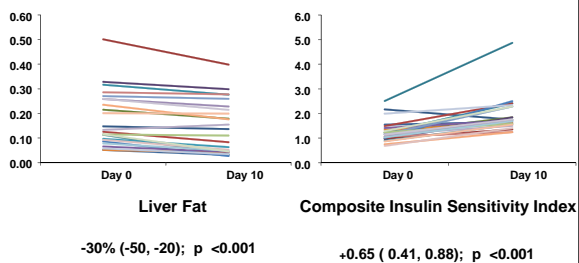
Oral glucose tolerance test before and after isocaloric fructose restriction



Lustig et al. Obesity (in press)

For pediatric subjects with
hepatic steatosis (n = 25)

Change in Liver Fat and Insulin Sensitivity



Lustig et al. Obesity Society. Nov. 4, 2015. Adjusted for change in weight

Correlation between Insulin Sensitivity & Liver Fat vs Visceral Fat

| Spearman R | Day 0 | | Day 10 | | Change in fat (Absolute) | |
|---------------------------------------|--------------------|--------------|--------------------|--------------|--------------------------|--------------|
| | Liver Fat Fraction | Visceral Fat | Liver Fat Fraction | Visceral Fat | Liver Fat Fraction | Visceral Fat |
| Insulin Sensitivity (CISI) DAY 0 | - 0.36 § | - 0.57* | | | | |
| Insulin Sensitivity (CISI) DAY 10 | | | - 0.28 | - 0.34 § | | |
| Change in Insulin sensitivity (ΔCISI) | | | | | - 0.56* | 0.09 |

* p < 0.05

§ 0.05 < p < 0.1

Lustig et al. Obesity Society, Nov. 4, 2015

Association between Insulin Sensitivity (CISI) and Liver Fat vs. Visceral Fat

Liver Fat

• Every unit decrease in liver fat increases CISI by 10.49 (p = 0.008).

- Every unit decrease in %liver fat increases %CISI by 0.8 (p = 0.03).

Visceral Fat

• Every unit decrease in visceral fat is non-significantly increases CISI by 0.002 (p = 0.77).

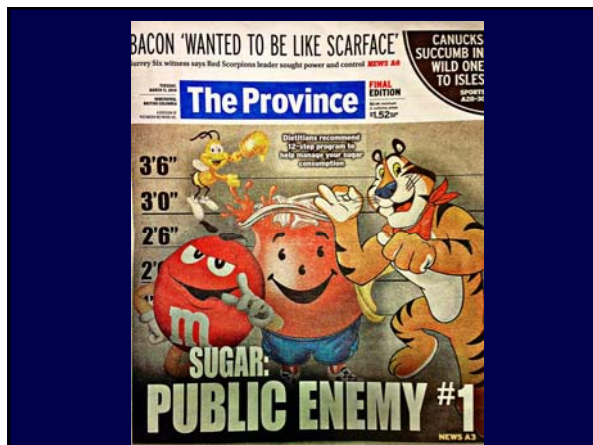
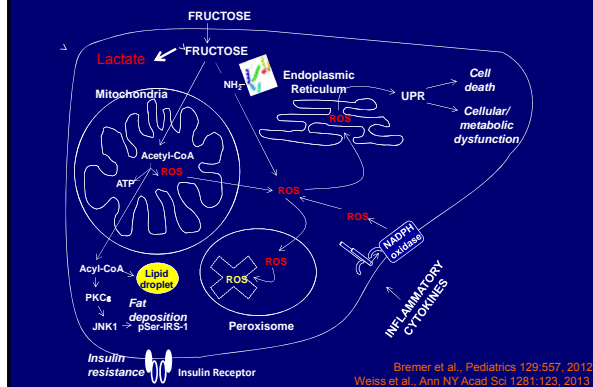
• Every unit decrease in %visceral fat non-significantly increases %CISI by 0.69 (p = 0.28).

Lustig et al. Obesity Society, Nov. 4, 2015

Summary

- **Prospective correlational** data demonstrate **associations** between added sugar and heart disease, **exclusive of calories or obesity**
- **Econometric** data show **causal medical inference** for added sugar and diabetes, **exclusive of calories or obesity**
- **Interventional isocaloric glucose for fructose** study shows improvements in fatty liver disease, insulin resistance and metabolic health in children in 10 days, and insulin resistance is driven by liver fat

Toward a unifying hypothesis of metabolic syndrome



Collaborators

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Stanford Prevention Institute

Sanjay Basu, M.D., Ph.D.