# Target discovery in metabolic disease

### Cord F. Dohrmann

The prevalence of metabolic diseases is taking on epidemic proportions and poses a serious threat to human health. Current treatment options have proven insufficient to cope with obesity and diabetes because they rarely restore normal metabolism and thus leave patients exposed to lifethreatening complications. Successful management of these diseases depends on novel, improved therapeutic strategies targeting early intervention in disease progression. Discovery of novel metabolic disease targets has been hampered by the complexity of contributing environmental and genetic factors, as well as the need for potent but safe treatments suitable for chronic diseases. Genomic approaches are excellent tools to manage genetic complexity and have been applied successfully to identify candidate target genes that will lead to the development of novel therapies for metabolic diseases.

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Metabolic diseases affecting energy homeostasis such as obesity, type 2 diabetes (T2D) and the metabolic syndrome are a major challenge for healthcare systems worldwide because they are strongly associated with several major health risk factors [1-4]. Currently marketed drugs, which are aimed at achieving long-term control of body weight and blood glucose levels, largely target symptoms with limited efficacy and considerable side effects that raise compliance issues [5,6]. Development of more effective drugs for metabolic diseases requires the identification of mechanisms and targets that have the potential to be developed into improved therapeutic but also preventative strategies.

Sequencing of the human genome combined with technological development of large-scale genomic approaches has made it possible to systematically investigate genetic factors for their involvement in human diseases. Genomic approaches such as large-scale gene expression analysis, functional screens in model organisms, and human genome scans for susceptibility genes have led to the identification of countless candidate drug targets for most major indications including metabolic diseases. Owing to the wide-ranging

technological spectrum of these approaches, there are tremendous differences in resources and timelines they require as well as the results they have produced so far. This review focuses on the contributions of these genomic technologies to target discovery in obesity and T2D, with an emphasis on applied strategies as well as emerging principles rather than individual results.

### **Expression profiling**

So far, expression-profiling efforts have relied to a large extent on technologies analyzing the composition of complex mRNA samples and, less frequently, protein samples. Widely used approaches for assessing the composition of mRNA pools or differences between them are differential display, subtractive cloning or, with increasing popularity, DNA microarrays carrying probe sets that cover essentially complete genomes [7]. Tissues and organs that are of particular interest in metabolic diseases include adipose tissue, muscle, liver, pancreas and the hypothalamus, which have pivotal roles in the regulation of energy storage, maintenance of blood glucose homeostasis and appetite control. Below, selected expression profiling studies are presented to exemplify how these technologies have been applied to address peripheral mechanisms of energy homeostasis, which could provide the underlying cause for several metabolic diseases (Table 1).

### Adipose tissue

Adipose tissue is not only the primary organ responsible for storage of excess energy but also fulfils important endocrine functions that affect peripheral and central mechanisms involved in maintaining energy homeostasis. First attempts to define the composition of adipose tissue on the mRNA and protein level have been carried out by Gabrielsson et al. [8] and Sanchez et al. [9], respectively. Further experiments investigated molecular differences

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| Experimental details  | Species        | Technology               | Refs               |
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| Adipose tissue  |                |                          |                    |
| Abdominal subcutaneous adipose tissue from non-obese individuals  | Human          | Microarray               | [8]                |
| White and brown adipose tissue from metabolically healthy mice  | Mouse          | Proteomics               | [ <mark>9</mark> ] |
| Perirenal visceral versus subcutaneous adipose tissue   | Rat            | Microarray               | [10]               |
| Abdominal omental adipose tissue from non-obese individual (female)   | Human          | Microarray               | [11]               |
| Omental versus subcutaneous from obese individuals (male)   | Human          | Microarray               | [12]               |
| Visceral versus subcutaneous adipose tissue from obese individuals (male/female)  | Human          | Differential<br>display  | [13]               |
| White adipose tissue from lean versus HFD induced obese mice  | Mouse          | Microarray               | [16]               |
| Adipose tissue from rats at the onset (7, 14, 21 days) of HFD-induced obesity   | Rat            | Subtractive cloning      | [15]               |
| Adipose tissue from genetically obese (ob/ob) versus HFD-induced obese mice   | Mouse          | Microarray               | [21]               |
| Adipose tissue from ob/ob versus HFD-induced obese mice   | Mouse          | Micorarray               | [22]               |
| Adipose tissue from lean versus ob/ob mice and leptin-treated versus food restricted  | Mouse          | Microarray               | [18]               |
| Adipose tissue from lean, obese and obese-diabetic mice   | Mouse          | Microarray               | [17]               |
| Omental adipose tissue from lean versus normotensive, normoglycemic obese individuals   | Human          | Microarray               | [19]               |
| Adipose tissue non-diabetic, insulin resistant versus diabetic individuals  | Human          | Microarrays              | [20]               |
| White adipose tissue versus brown adipose tissue  | Rat            | Microarray               | [31]               |
| Adipocytes  |                |                          |                    |
| 3T3-L1 preadipocytes (day 0) versus differentiated adipocytes (day 6)   | Mouse          | Micorarray               | [25]               |
| 3T3-L1 in vitro differentiation time course (24 hours, 4 days, 1 week)  | Mouse          | Micorarray               | [26]               |
| 3T3-L1 early in vitro differentiation time course (0, 2, 8, 16, 24 hours)   | Mouse          | Microarray               | [27]               |
| 3T3-L1 differentiation time course versus primary predipocytes and adipocytes from WAT  | Mouse          | Microarray               | [28]               |
| Primary preadipocytes versus adipocytes from subcutaneous adipose tissue (average BMI 29)   | Human          | Microarray               | [29]               |
| NIH-3T3 fibroblasts stably transfected with PPAR- $\gamma$ versus vector transfected  | Mouse          | Subtractive cloning      | [50]               |
| 3T3-L1 preadipocytes differentiated in the presence of PPAR-γ agonist (rosiglitazone)   | Mouse          | Differential<br>display  | [49]               |
| Muscle Muscle tissue from ST7 treated (insulin deficient) dishetic miss   | Mayraa         | Microconov               | [22]               |
| Muscle tissue from STZ-treated (insulin deficient) diabetic mice  | Mouse          | Microarray               | [33]               |
| Gastrocnemius muscle from metabolically healthy mice  | Mouse          | Proteomics               | [9]                |
| Muscle tissue from human skeletal muscle during hyperinsulinemic clamp  Skeletal muscle of type 2 diabetes patients during intense insulin treatment versus untreated | Human<br>Human | Microarray<br>Microarray | [34]<br>[35]       |
| Muscle of type 2 diabetic patients (insulin-resistant) versus non-diabetic (insulin-sensitive)  | Human          | Microarray               | [36]               |
| Muscle from non-diabetic, insulin-resistant versus non-diabetic, insulin-sensitive  | Human          | Differential<br>display  | [38]               |
| Muscle from obese non-diabetic, insulin-resistant versus non-diabetic, insulin-sensitive Liver  | Human          | Micorarray               | [37]               |
| Liver tissue from metabolically healthy mice  | Mouse          | Proteomics               | [9]                |
| Liver tissue from obese diabetes-susceptible versus obese but diabetes-resistant mice   | Mouse          | Microarray               | [42]               |
| Liver tissue from STZ-treated diabetic mice   | Mouse          | Microarray               | [41]               |

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| Experimental details  | Species | Technology     | Refs                |
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| Liver (continued)   |         |                |                     |
| Liver tissue from leptin-deficient mice, leptin-treated versus untreated                | Mouse   | Microarray     | [ <del>40</del> ]   |
| Liver tissue from lean and obese mice treated with PPAR- $lpha$ and - $\gamma$ agonists | Mouse   | Proteomics     | [ <mark>52</mark> ] |
| Liver tissue from rats treated with PPAR-γ agonist                                      | Rat     | GeneCalling    | [51]                |
| Pancreas  |         |                |                     |
| Isolated pancreatic islets from metabolically healthy mice                              | Mouse   | Proteomics     | [ <mark>9</mark> ]  |
| Developmental stages of pancreas development, isolated islets, insulinoma               | Human   | EST sequencing | [7]                 |
| Fetal pancreas, isolated islets and insulinoma  | Mouse   | EST sequencing | [ <mark>7</mark> ]  |
| Pancreatic tissue from partially ligated pancreatic duct mice (pancreas regeneration    | Mouse   | Microarray     | [49]                |
| model)  |         |                |                     |
| Isolated and cultured human islets  | Human   | Microarray     | [48]                |

Abbreviations: BMI, body mass index; HFD, high fat diet; STZ, streptozotocin; WAT, white adipose tissue.

between subcutaneous and visceral human adipose tissue [10-13]. The results indicate that adipose tissue is more complex than originally appreciated. In addition to simply storing and mobilizing lipids, adipose tissues produce and release metabolically active polypeptides or metabolites that can be involved in immunity (complement factors), endocrine function (leptin), metabolism (adiponectin, resistin), and cardiovascular control (angiotensinogen, plasminogen activator inhibitor-1) [14]. Furthermore, experiments investigating molecular differences between subcutaneous and visceral human adipose tissue strongly suggest that fat depots from various regions of the body are functionally diverse in the way they respond to endocrine factors, the adipokines they produce and whether or not they are able to carry out thermogenesis [10–13].

Other studies addressed changes in adipose tissue gene expression during the transition from a metabolically healthy state to increasingly diseased states. Li et al. [15] determined that genes involved in the expansion of nonadipose tissue are upregulated at the onset of diet-induced obesity (DIO) possibly driving the proliferation of preadipocytes. By contrast, Moraes et al. [16] looking at later stages of DIO, found genes involved in lipid metabolism and adipogenesis to be downregulated and inflammatory markers to be upregulated. Specifically, the observation that genes normally associated with adipocyte differentiation are downregulated in the obese state is supported by studies in genetically obese mice, strongly suggesting that adipose tissue of obese mice might not be fully functional and therefore contribute to the progression of the disease from an obese but non-diabetic state to the obese-diabetic state [17,18].

More recent studies carried out in humans comparing the insulin resistant but non-diabetic versus the diabetic state and the lean versus the obese state also provide evidence that there are significant changes in gene expression in diseased adipose tissue. For example, variations in gene expression in omental adipose tissue from normoglycemic, normotensive obese patients (BMI 37.3 kg m<sup>-2</sup>) when compared to lean individuals (BMI 23.4 kg m<sup>-2</sup>) are associated with a broad spectrum of biological processes. In obese patients, these changes include an increase in the expression of genes involved in lipolysis repression, mitogen-activated protein kinase signalling, angiogenesis and immune function, whereas there is a decrease in the expression of genes involved in lipolysis activation, preadipocyte proliferation and differentiation. Similarly, in the insulin resistant state, early and late markers of the adipogenic process are downregulated, lending further support to the hypothesis that adipose tissue in obese patients is functionally impaired [19,20]. Obesity is associated with a state of chronic, lowgrade inflammation that appears to reside predominantly in adipose tissue and is characterized by macrophage infiltration. It is hypothesized that macrophages in adipose tissue might contribute to the elevation in circulating inflammatory markers, which have been linked to the development of insulin resistance [21,22]. Regardless of whether or not the initial inflammatory response emerges from the adipocyte and further propagates with the recruitment of macrophages or whether macrophages infiltrate the adipose tissue and subsequently initiate the inflammatory response themselves, treating inflammation of adipose tissue could provide an alternative mechanism for pharmacological intervention [23].

Targeting adipogenesis is a proven mechanism for improving insulin resistance, as demonstrated by thiazolidinediones (TZDs) such as Avandia (GlaxoSmithKline; http://www.gsk.com) and Actos (Lilly; http://www.lilly.com),

which are agonists of peroxisome proliferator-activated receptor (PPAR)-y, a key driver of adipocyte differentiation and an important drug target for T2D. Identification of genes in related pathways with similar effects on adipogenesis or adipocyte function might provide alternative targets without the potentially target-related side effects observed with TZDs. The process of adipogenesis has been studied predominantly in the mouse 3T3-L1 preadipocyte cell line [24]. Various groups have analyzed gene expression during adipogenesis looking either at early or late stages of the differentiation process [25–27]. These studies revealed many functionally uncharacterized genes that are regulated during adipogenesis as well as differences in gene expression between human and mouse adipogenesis previously not appreciated [28,29]. It is of interest to note that in addition to PPAR-y, several genes that are upregulated during adipogenesis are being pursued as potential drug targets. These include 11-β-hydroxysteroid dehydrogenase type-1 (11-β-HSD-1), probably the most advanced target in regards to drug development, but also fatty acid binding protein, hormone-sensitive lipase (HSL), and resistin, among others.

Another attractive mechanism, especially for the treatment of obesity, is to increase energy expenditure through activation of uncoupling proteins (UCPs). Whereas the physiological role of white adipose tissue (WAT) is to store excess energy in the form of fat, the main function of brown adipose tissue (BAT) is to use existing fat stores for thermogenesis via activation of UCP1 [30]. Identifying genes that are essential for efficient energy dissipation in BAT will lead to a better understanding of UCP-related pathways and components of these pathways could provide potential metabolic disease targets. Unami *et al.* [31] addressed this question through gene expression analysis of WAT and BAT identifying many differentially expressed genes which await further functional analysis.

### Muscle

Similar to adipose tissue, skeletal muscle has a central role in metabolic diseases primarily because it is the major site of glucose disposal in response to insulin and thus the main site of insulin resistance in T2D. Various studies have investigated the transcriptional response to an insulin stimulus, identifying >150 insulin-regulated genes [32]. More recent *in vivo* studies analyzed the changes in gene expression in skeletal muscle of mice either lacking an insulin signal, during hyperinsulinemic euglycemic clamp, or untreated versus insulin-treated T2D [33–35]. Furthermore, in attempts to identify molecular mechanisms, and thus candidate drug targets, associated with insulin resistance in skeletal muscle, gene expression analysis was conducted in non-diabetic versus diabetic

and insulin sensitive versus insulin resistant but non-diabetic human subjects [36–38]. All of these studies indicate that insulin action, or the lack thereof in skeletal muscle, leads to highly coordinated changes in gene expression in various metabolic pathways, including glucose, protein and lipid metabolism but also cell proliferation and survival. Despite the fact that the insulin pathway has been extensively investigated and many of its components are pursued in drug discovery efforts (e.g. protein tyrosine phosphatase-1b, glycogen synthase kinase-3, protein kinase B, SH2 domain-containing inositol 5-phosphatase), the complex intracellular signalling network involved offers many more targets with the potential to attack insulin resistance directly.

### Liver

The liver has a central role in maintaining blood sugar levels by acting as a storage depot during periods of dietary carbohydrate excess and as a glucose source when carbohydrates are unavailable. Hepatic glucose production is inappropriately high in T2D primarily due to increased gluconeogenesis rather than glycogenolysis. The biguanide Metformin [Merck-Lipha (http://www.lipha.fr) and Bristol-Myers Squibb (http://www.bms.com)] is the only marketed agent targeting hepatic glucose production as its primary mode-of-action. Other protein targets that control hepatic glucose production that are currently being pursued (such as glucagon receptor, glycogen phosphorylase, glucocorticoid receptor, fructose-1,6-bisphosphatase, carnitine palmitovltransferase-1, glycogen synthase kinase-3, glucose-6-phosphate translocase, adenosine A2B receptor and 11-β-HSD-1) are still in preclinical and clinical stages of investigation [39]. Further candidate target genes regulating the coordinated expression of key gluconeogenic and glycogenolytic enzymes, as well as other mechanisms, should certainly be among those genes identified in expression profiling efforts investigating the liver under metabolic stress, such as absence of an insulin stimulus or in the diabetic or obese state [40–42]. These studies revealed that a wide variety of genes involved in fatty acid metabolism are transcriptionally regulated in the diseased state, suggesting that hepatic fat metabolism might also offer potential targets for intervention.

### Pancreas

Another important peripheral organ for the treatment of T2D is the pancreas, where primarily endocrine hormone-producing islets are targeted to secrete more insulin in response to elevated blood glucose levels. Although insulin secretagogues, including sulfonylureas, glitinides and meglitinides, are important therapeutic options, it is likely

that future drugs will target the preservation of β-cell function and β-cell mass rather than insulin secretion. In particular, the concept of targeting β-cells directly has received much support from work by Dor et al. [43], suggesting that terminally differentiated β-cells retain significant proliferative capacity in mice and have a significant role in maintaining β-cell mass. Glucagon-like peptide (GLP)-1 analogues such as exenatide (Eli Lilly) are in late stage clinical trials and have been shown to improve β-cell function and potentially even increase β-cell mass, making it likely that more targets that are directed at similar mechanisms will follow [44]. Dipeptidyl peptidase (DPP) IV inhibitors, which are currently in advanced clinical development [Novartis (http://www.novartis.com) and Merck (http://www.merck.com)], target essentially the same mechanism by preventing the degradation of endogenous GLP-1 [45]. Similar to the GLP-1 analogues, there is evidence indicating that a lack of DPP IV function might prevent streptozotocin-induced as well as high fat diet-induced β-cell apoptosis and potentially might even contribute to islet neogenesis [46,47]. Molecular drug target candidates are rare in this area and published reports about expression profiling approaches aimed at the identification of pancreatic drug targets are limited [7,9,48,49].

### Expression profiling of disease-relevant pathways

Alternative target discovery approaches are aimed at the elucidation and dissection of highly disease-relevant pathways such as for PPAR-y, leptin and insulin. Pharmacological treatment, genetic loss or gain of function studies in appropriate cell lines or tissues combined with differential gene expression analysis can lead to the identification of transcriptionally regulated components of specific signalling pathways, which might be suitable for therapeutic intervention. An example for studying insulin action was already presented in the study by Rome et al. [34]. PPAR- $\gamma$  and - $\alpha$  agonists have been used in vitro and in vivo to elucidate ligand-induced changes in mRNA expression in adipose tissue and liver [50–53]. Although the main intention was to evaluate compound efficacy and potentially identify useful clinical markers, these studies also revealed unexpected changes in expression patterns, which could lead to a more comprehensive picture of these pathways. Finally, the leptin pathway is of interest because of the potent effects of leptin on food intake, energy expenditure and many other metabolic processes. Soukas et al. [18] administered leptin to wild-type and leptin-deficient mice to identify kinetically distinguishable gene clusters that are specifically regulated by leptin. The results strongly suggest that in addition to its central effect on food intake, leptin activates metabolic pathways that deplete adipose tissue, making the leptin signal transduction pathway a highly interesting target in peripheral tissues as well as the CNS. Thus, despite the fact that clinical trials with leptin have been disappointing in regard to energy expenditure and weight loss potentially due to an underlying leptin resistance in obese patients, components of the leptin signal transduction pathway should prove interesting drug targets in the future.

### The limits of expression profiling

Most of the expression profiling experiments described above have led to the discovery of large numbers of regulated genes and gene clusters and therefore represent an excellent first step towards the identification of mechanisms associated with a certain physiological state, and thus potential metabolic disease targets. However, it is important to keep in mind that expression-profiling approaches have technical and conceptual limitations. Although the technologies are constantly improving, the majority of the presented studies provide neither a comprehensive picture regarding the transcriptional status of the complete genome nor an adequate representation of the dynamic range of transcriptional regulation. It remains a challenge, particularly for microarray-based studies, to decide at what point a gene should be classified as transcriptionally regulated because the dynamic range of detection is usually limited. It is also important to note that observed changes on the transcriptional level are not necessarily reflected on the protein level and that genes whose activity is primarily regulated on the post-transcriptional level will not be detected by RNA-based technologies. Furthermore, for all studies comparing the expression status of healthy and diseased tissues, it remains to be determined if transcriptional regulation of a particular gene is directly related to a primary defect or is rather the consequence of a prolonged pathophysiological state. For example, in patients with increased free-fatty-acid levels, the primary defect could be an increase in lipolysis, which in turn causes insulin resistance, or alternatively the primary defect could be insulin resistance, which leads to an increase in lipolysis. Such questions are probably addressed most easily in systems where it is possible to follow disease development and progression in detail, such as model organisms. Despite these issues, expression-profiling studies have provided a wealth of candidate genes, which have been associated with particular physiological functions or disease mechanisms. The future challenge will be to mine these data appropriately and select the most interesting genes for further functional validation. In summary, expression-profiling studies are highly valid approaches to identify novel mechanisms involved in metabolic diseases as well as potential candidate target genes as demonstrated by several examples. However, in most cases they only represent a first step that requires in depth analysis of the generated datasets, as well as additional functional experiments to select high potential targets for further drug discovery efforts.

### **Human genome scans**

Genome-wide scans for susceptibility genes are conducted with the use of genetic polymorphisms such as simple tandem repeats or single nucleotide polymorphisms, performing linkage analysis in usually related individuals with a certain disease and/or characteristic quantifiable trait associated with the disease. A large number of human genome scans have been performed and there are comprehensive reports detailing the results for susceptibility loci associated with T2D, obesity and the metabolic syndrome [54–56].

The emerging picture is one of high complexity, with potential susceptibility loci distributed all over the human genome except for the Y chromosome. Interestingly, several independent studies provide accumulating evidence that a significant number of loci are being linked to both T2D and obesity, strongly suggesting a common underlying genetic architecture [57,58]. Due to the inherent genetic complexity of multi-factorial diseases, the contributions of individual gene variants are usually modest and therefore hard to detect, especially if there is variation among analyzed patient populations and the screened patient numbers are relatively small. The problem of genetic complexity is compounded by the tremendous variability of encountered environmental factors (e.g. diet, job, exercise). Nevertheless, there is accumulating evidence for significant association of ~90 candidate genes with obesity. However, in many cases linkage only accounts for a small fraction of the patient population and in some cases might even turn out to be false positives. In the case of T2D, overall >25 susceptibility genes have been identified, at least 21 of which are responsible for maturity-onset diabetes of the young (MODY), now referred to as the monogenic and/or syndromic form, whereas only six genes have been associated with multi-factorial T2D. In monogenic T2D the primary defect is in insulin secretion and not adiposity combined with peripheral insulin resistance, which are hallmarks of multi-factorial T2D. The pathophysiological basis for multi-factorial T2D is not well established but clearly important contributing factors are peripheral insulin resistance coupled with a progressive impairment of insulin secretion. The progressive loss in  $\beta$ -cell function is becoming increasingly important for multi-factorial T2D, therefore, the borders between the monogenic and multi-factorial T2D are less well defined. Currently, there are six different MODY genes, accounting for 90% of MODY families, which predominantly affect insulin secretion and have been identified through a combination of positional cloning and candidate gene-based analyses. Overall, despite significant linkage of many loci with metabolic diseases, the only susceptibility gene identified by positional cloning efforts alone is clapain-10 (CAPN-10) for multi-factorial diabetes [59].

So far, in comparison to human genome scans, it has been more productive to screen patients for mutations in candidate genes that were initially identified by other means. This 'candidate gene' approach has been reasonably successful in linking mutations in at least 15 genes (β adrenergic receptors 2 and 3, guanine nucleotide binding protein, leptin receptor, leptin, melanocortin 4 receptor, glucocorticoid receptor, proprotein convertase subtilisin/kexin type 1, proopimelanocortin, PPAR-γ, PPAR-γ coactivator 1, protein phosphatase 1 regulatory subunit 3A, tumor necrosis factor, single-minded homolog 1, and uncoupling proteins 1-3) to obesity, as a dominant clinical feature. However, considering that several hundreds of biological candidate genes, previously implicated to be involved in β-cell, adipocyte, liver or other metabolic function, have been examined for their potential role in metabolic disease these are relatively few cases where good evidence for a contribution to the susceptibility of metabolic disease could be established [55,59]. In summary, genome scans have not contributed significantly to the identification of novel targets so far but are a valuable resource for confirming the relevance of target gene candidates. The combination of positional cloning efforts and association studies of candidate genes will deliver many more metabolic disease susceptibility genes, such as CAPN-10, and thereby significantly improve our understanding of the pathophysiological basis of metabolic diseases, and ultimately provide novel mechanisms and molecules to be targeted.

## Functional screens for metabolic disease targets in model organisms

Functional screens in model organisms are a powerful approach to functionally assign genes to certain biological processes and are therefore well suited to identify key components of mechanisms controlling metabolic homeostasis. The first step usually involves mutagenesis of the genome via chemical, genetic or RNAi-based technologies, which is then followed by scoring of individual mutants or mutant lines for relevant phenotypes. Screens covering large portions of the genome have been performed successfully in fruit flies (*Drosophila melanogaster*), zebrafish and mice [60–62]. However, published reports of genome-wide

model organism screens specifically designed to identify metabolic disease phenotypes are still the exception although they are highly complementary to expression profiling and human association studies.

### Mouse

The mouse is the most important model organism for target validation as demonstrated by the fact that knockout phenotypes for the 100 best-selling drugs correlate well with known drug efficacy [63]. This paradigm seems to hold true for metabolic diseases and is reflected in pipeline targets of the pharmaceutical industry (serotonin receptor 5-HT2C, cannabinoid CB1 receptor, cholecystokinin receptor, GLP-1 receptor, DPP IV) in the area of obesity and T2D that are currently in late stage clinical trials [64]. Although various approaches to undertake systematic genome-wide functional screens in the mouse have been initiated, none of these screens was specifically focused on metabolic phenotypes. Generally, mouse models with metabolic phenotypes have been generated by selective targeting of a particular gene, which was either previously implicated in metabolism or has been generated with unrelated objectives/indications in mind. The only exceptions are naturally occurring mouse mutants, such as leptin-deficient ob/ob and tubby mice, where the underlying mutations have been identified by positional cloning efforts [65,66].

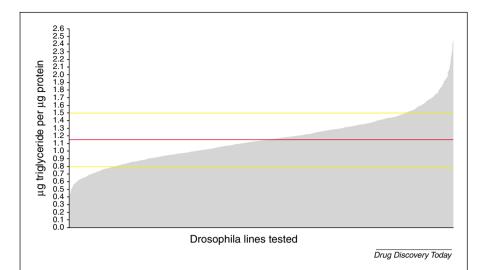
### Caenorhabditis elegans

The worm is not necessarily an organism that comes to mind when thinking about metabolic diseases, nevertheless, at the present time only the genome of Caenorhabditis elegans has been subjected to a nearly comprehensive screen specifically for a lipid storage phenotype. Many human disease-associated genes are present in the C. elegans genome and metabolic processes are evolutionary highly conserved, therefore it seems reasonable to use the worm for the identification of metabolic disease targets [67]. Ashrafi et al. [68] knocked down the function of 16,757 worm genes by applying RNAi covering nearly the complete genome while screening for lipid storage phenotypes. Although worms lack proper adipose tissue, they have specialized intestinal cells that store lipids very effectively. Considering that obesity and the metabolic syndrome are diseases characterized by improper storage of lipids, it is probably not too surprising that several genes with known metabolic functions were identified that validate the approach. Overall, the screen yielded 305 genes that caused a reduction in fat storage when inactivated and 112 genes that caused an increase in fat storage. Depending on the level of stringency applied, it is possible to assign human homologues to 30-50% of the identified worm genes. According to the authors, most of these genes have not been previously implicated in regulating fat storage and therefore could provide novel candidate targets for obesity and associated diseases.

### Drosophila melanogaster

An initial analysis of the fruit fly genome for probable orthologues of human disease genes identified matches to 75% of the 1378 human disease loci, indicating that flies are more closely related to humans than worms. In particular, genes implicated in metabolic disorders are almost always conserved, with the insulin receptor signaling pathway being a prominent example [69]. Flies have abdominal adipose-like tissue, which is morphologically and physiologically similar to mammalian adipose tissue. Doane [70,71] even identified a naturally occurring mutant fly line with an obesity-related phenotype arising from a mutation in a protein of unknown function, which combines WD40 and tetratricopeptide protein-protein interaction motifs [72]. Furthermore, it has been reported that Drosophila has specialized neuronal cells that produce insulin and are functionally orthologous to mammalian β-cells. Ablation of these insulin-producing cells in flies generates phenotypes similar to diabetes [73].

The basic metabolic processes regulating fat as well as carbohydrate metabolism in Drosophila are conserved and the tools for genetic screens are well established, therefore, researchers at DeveloGen (http://www.develogen.com) set out to screen the Drosophila genome for metabolic phenotypes. A screen for modifiers of UCP activity yielded genes involved in the regulation of UCP activity, exemplified by an inner mitochondrial transmembrane protein, which suppresses UCP activity in flies as well as in mammals [74]. A second screen determining the overall triglyceride content of >10,000 mutant Drosophila lines, demonstrates that it is possible to detect fly lines with either significantly increased triglyceride levels (obese), or significantly decreased triglyceride levels (lean; Figure 1). The fact that the triglyceride levels of the majority of fly lines is within 30% of the average strongly suggests that mutations in most genes have only a limited effect on overall triglyceride storage, whereas mutations in relatively few genes have a pronounced effect on triglyceride storage. Although this point might seem trivial, it nicely illustrates the power of functional screens in model organisms, where animals can be kept under standardized conditions to produce metabolic phenotypes based on mutations in genes that probably could not have been predicted. That predictions about gene functions are inherently difficult even for well characterized genes was demonstrated once again by example of HSL, which until recently was believed to be the rate-limiting



**Figure 1.** Overall triglyceride levels in mutant Drosophila lines were determined in relationship to protein content. The distribution of triglyceride levels reveals that most flies carrying a P element-induced mutation in one particular gene have an average triglyceride content of around 1.1  $\mu$ g triglyceride per  $\mu$ g protein (red line). Wildtype flies vary in their triglyceride content by about 30% (yellow lines), suggesting that variations beyond this threshold are significant. Displayed are triglyceride levels of a batch of 2108 fly lines analyzed under identical conditions.

enzyme for adipocyte lipolysis but turned out to be less than critical for triglyceride hydrolysis in HSL-deficient mice [75,76].

Many of the genes identified in these Drosophila screens are not only highly relevant to energy metabolism in flies but are clearly involved in mammalian metabolic processes, such as adipogenesis, fat or carbohydrate metabolism, and insulin receptor signalling. In addition, many genes that had not been previously implicated in the regulation of energy metabolism have also been identified and therefore represent novel candidate target genes.

### The potential of model organisms

Compared to expression profiling or human genome scans, functional screens in model organisms have several advantages when it comes to target discovery in metabolic diseases: (1) Metabolic diseases are of multi-factorial origin, caused by a complex interplay of the genetic predisposition of an individual and environmental factors; dissecting cause and consequence is only possible if all variables except for one are kept constant, which is usually not feasible in approaches involving human subjects. (2) Key metabolic processes, including appetite, fat and carbohydrate metabolism, as well as basic metabolic rate are governed by complex interwoven pathways that are characterized by considerable redundancy and a variety of feedback mechanisms: identification of genes that are either rate-limiting and/or lack redundancy is most easily accomplished

in whole-organism screens. (3) Drugs for metabolic diseases usually need to be administered chronically and therefore require excellent safety profiles, essentially devoid of target-related adverse effects; screening in model organism quickly reveals if inhibition of a certain gene function has the potential to generate unwanted adverse effects. Thus, functional screens for metabolic disease targets in model organisms still present an enormous and so far largely untapped potential for future target discovery efforts in metabolic diseases.

### Conclusion

Target discovery in the area of metabolic diseases is a field of intense investigation, and will be for years to come, because of the obvious ethical and financial challenges for human health and healthcare systems posed by metabolic diseases. Although highly interest-

ing targets are currently pursued in drug discovery efforts, the pathophysiological basis of metabolic diseases is only poorly understood. Identification of targets associated with the early pathophysiology of metabolic diseases could lead to the development of preventative therapeutic strategies suitable for earlier intervention than is possible with current treatments. Large-scale expression profiling, the search for human susceptibility genes, as well as model organism screens are highly complementary approaches for target discovery in metabolic disease. The combination of these genomic approaches will deliver novel insights regarding the etiology of T2D, obesity and metabolic syndrome and thus undoubtedly lead to the development of novel and improved therapies based on highly validated drug targets.

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